This document contains the point-to-point responses to the reviews #1, #2, #3 and to the short comments and editor comment, followed by a marked-up version of the manuscript and the supplemental material.

List of relevant major changes in the manuscript:

- Change of the title of the paper to avoid the misleading phrase Euler-Lagrangian descriptor (We have replaced the phrase Euler-Lagrangian descriptor in the whole manuscript by Lagrangian descriptor $M_V$ based on the modulus of vorticity)
- Correction of misleading statements in the introduction (pp. 2-3 in the marked up version) as well as including the references pointed out by the referees
- Extended explanation of the Lagrangian descriptors $M$ and $M_V$ and their features (Sect. 2 pp. 5-7 in the marked-up version)
- Change of the colorcode Fig. 1, Fig. 2, Fig. 3, Fig. 4, Fig. 5, Fig. 7, Fig. 8 and Fig. 9
- Extended explanation the idea of the eddy and eddy tracking based on $M_V$ (p. 8 line 16 to p. 9 line 19 in the marked-up version)
- Extended discussion of the choice of $\tau$ including an improved Fig. 6 (p. 10 line 22 to line 26 in the marked-up version)
- Partly rewritten Sect. 4.2 to clarify the concept of noise and what is aimed with the idea to add noise to the velocity field (Sect. 4.2 pp. 10-13 in the marked-up version)
- Rewritten Sect. 4.3 including a revised version of the eddy shape detection only based on $M_V$ and an example of the eddy shapes for the western Baltic Sea instead of the seeded eddy model example (Sect. 4.3 pp. 13-17 in the marked-up version)
- New Fig. 10 showing the results of the eddy shape detection based on $M_V$
- New Fig. 11 showing eddy shapes for an example in the western Baltic Sea
- Rewritten Sect. 5 Discussion and conclusion explaining the advantages and disadvantages of the Lagrangian descriptor $M_V$ and the eddy tracking based on it
- Minor changes in phrasing, grammar, spelling and punctuation in the whole manuscript as well as the correction of equation 3 (p. 5 line 20 in the marked-up version)

List of the relevant major changes in the supplemental material:

- Deleted Sect. S1 about the seeded eddy model, because we have replaced the example in the manuscript (p. 1 in the marked-up version)
- Extended Section about the algorithm of the eddy tracking. Now it includes an explanation of the eddy shape detection based on $M_V$ and a new Fig. S3 (pp. 2-6 in the marked-up version)
- Minor changes in phrasing, grammar, spelling and punctuation in the whole supplemental material
Response to Reviewer #1

We would like to thank the reviewer for his critical assessment of our manuscript and the suggestions to improve it, which we have taken into account. In the following we respond to all the concerns of the reviewer and indicate the changes in the manuscript:

The authors develop a new Lagrangian descriptor based on integrating the magnitude of the vorticity (an Eulerian quantity) along trajectories. They claim this is a new Eulerian-Lagrangian descriptor”. However, their approach follows exactly the methodology for Lagrangian descriptors described in the following paper:
At the bottom of page 3532 of this paper it is stated that Lagrangian descriptors can be constructed by integrating any bounded positive intrinsic physical or geometric property of the velocity field...". Certainly the magnitude of the vorticity satisfies that criteria, but also the magnitude of the velocity field, which is a common Lagrangian descriptor used in the abovementioned paper (henceforth, Mancho et al 2013). It just so happens that the integral of the square root of the magnitude of the velocity field along trajectories has the interpretation of arclength, but it is still of the same character as the quantity studied by the authors who wish to rename the quantity Eulerian-Lagrangian descriptor”. This is completely misleading and contrary to the methodology introduced in Mancho et al 2013. Indeed, all of the quantities in Mancho et al. 2013 would then be “Eulerian-Lagrangian descriptors” in the terminology of the authors of the paper under review as all of the quantities of Table 1 in Mancho et al. 2013 proposed for the construction of Lagrangian descriptors are Eulerian quantities. In this sense all Lagrangian descriptors are constructed from Eulerian and Lagrangian quantities, with the purpose of providing Lagrangian transport information. So there is nothing new methodologically" in this paper.

We completely agree with the reviewer and also reviewer #3 that already Mancho et al. 2013 pointed out, that any fluid property can be used to construct a Lagrangian descriptor. This has been pointed out by us explicitly already in the first version of the manuscript (cf. the sentence “As already pointed out by Mancho et al. (2013) any intrinsic physical ...” in the beginning of the paragraph before formula (4)). Our motivation to introduce the name Euler-Lagrangian descriptor was a more practical one. Because we found it difficult to read talking about one and another Lagrangian descriptor, we introduced a distinction by the names Euler-Lagrangian for one of them and Lagrangian for the other. Since this has been found misleading by two reviewers (#1 and #3) because it would look like the definition of a new descriptor, which indeed is not the case as we of course know, we have changed this in the revised version. To emphasize this we have now avoided the name Euler-Lagrangian descriptor in the title and throughout the whole manuscript. Furthermore, we have even more than before emphasized that the original idea had been already formulated in Mancho et al. 2013. We now write “We would like to emphasize, that it has been already pointed out by Mancho et al. (2013)”.

The authors claim that Lagrangian descriptors are not objective, and justify this claim by referring to reference [15]. However, we have looked at reference [15], and I cannot find any proof of non-objectivity for Lagrangian descriptors in that reference. If the authors are going
to make such a strong claim, then they must provide a reference to a proof of their claim.

We have removed the word heuristic as well as the whole discussion about objectivity from the text to avoid any further discussion of this issue since it is not the aim of this manuscript to clarify a question, which is debated in the literature. Our aim is much more practical and does not claim to develop a new mathematical method. Since the reviewer does not find Ref. [15] appropriate, we are thankful to George Haller to provide his comment #2 containing the arguments requested. Ana Mancho has answered the question of objectivity of Lagrangian descriptors in an editor’s comment. The main focus of this manuscript is completely different and does not concentrate on the question whether the method is objective or not. It is just an application of an existing method to tackle the question of providing a robust method to detect and count eddies in an oceanographic flow. We regret that the problem of objectivity distracted the reviewer from the main focus of the manuscript. However, since this problem has been addressed in several comments, we cannot ignore it and refer the reader now to this discussion in the discussion section of the journal.

The authors are proposing what they refer to as a new characterization of eddies based on an elliptic region bounded by segments of stable and unstable manifolds of a hyperbolic trajectory. This allows the lobe dynamics mechanism to control transport in and out of the elliptic region. However, a careful development of eddies from this point of view has already been given in the following references:


We would like to thank the reviewer for pointing out those papers to us, which we now cite in the text.

The authors claim that Lagrangian descriptors cannot detect eddies. I find this to be a very surprising statement based on several papers in the literature that characterize eddies in the Gulf stream and Gulf of Mexico, for example, in terms of Lagrangian descriptors.

We did not claim in the manuscript that Lagrangian descriptors cannot detect eddies. We have only written that the Lagrangian descriptor using the path length identifies both the eddy core and the DHTs with a minimum of M. This is a bit cumbersome when designing an algorithm for which it is impossible to check for each whether it corresponds to a hyperbolic point or elliptic point (eddy core). Therefore, we were looking for another Lagrangian descriptor not posing this difficulty. Using the Lagrangian descriptor Mv has the advantage that no distinction is needed between eddy cores and DHTs, since they are displayed as maxima and minima of Mv respectively. By contrast, the Lagrangian descriptor M needs an additional criterion since eddy cores and DHTs are both displayed as minima. Since the reviewer misunderstood our statements we have now reformulated it in a more precise way.

The word "heuristic" and the phrase "to identify Lagrangian coherent structures in a flow" are used very bizarrely, and incorrectly, here. First, there is nothing "heuristic" about this approach. A hyperbolic trajectory is a trajectory of the fluid flow having stable and unstable manifolds. The stable and unstable manifolds are made up of trajectories, this is why trajectories cannot cross them. They ARE ow barriers by construction (they do not have to be
"identified"). In other words, they are the direct construction of Lagrangian coherent structures. FTLEs and Lagrangian descriptors are methods to detect these structures (not to construct them).

According to our previous answer to the same criticism above we have deleted the word heuristic.

The word “identified” was used in the meaning that the manifolds are visible in the plot of M. We apologize for the misleading use of the word. We are not native speakers.

In several places the authors use the word "ridges" to refer to some property of Lagrangian descriptors. It is not clear what they mean by this since, to my knowledge, it has not been used in the literature to refer to any property of Lagrangian descriptors. It was used in the original Shadden and Marsden paper to refer to a feature of FTLE fields, and in that paper it was given a precise mathematical meaning. However, that meaning appears to have largely been "lost" as people now tend to throw around the term rather cavalierly (as these authors have done) without providing an understanding of its context and mathematical definition in the situation in which they are writing.

Indeed, as the reviewer pointed out, the word „ridges“ has been used in the context of FTLE fields, not only in the paper by Shadden and Marsden, but also in the papers by the group of Hernandez-Garcia/Lopez and coworkers from Spain. It just describes the local maxima of a certain quantity under investigation. We have deleted the word ridges in the context of Lagrangian descriptors.

In Section 4.2 the authors add noise to their velocity field. The details of this are not clear, especially if, and how, Lagrangian descriptors would fit into this framework. Lagrangian descriptors are integrations of positive quantities over trajectories. Noisy velocity fields give rise to stochastic ODEs, whose solutions are stochastic processes, not trajectories.

The reviewer is correct in saying that one has to study stochastic differential equations when dealing with noisy velocity fields. This would apply if one would be interested in a stochastic view of the problem. However, in the manuscript we look at this problem from a very different point of view. Since we are interested to design an algorithm searching and detecting eddies in a real velocity field from oceanography, we wanted to test the algorithm against noisy velocity field in the way, that the noise is observational noise or in other words measurement errors. For many problems, the velocity field will not be given as solutions of a numerical integration but from observational data, which are corrupted by measurement errors. Therefore, we just added noise to the computed velocity field. This approach enables us to “mimic” in a simple way a noise, which comes from errors in field measurements. This is at the same time a test, how a Lagrangian descriptor would respond to velocity fields which do not fulfill the mathematical criteria of a two-dimensional divergence free velocity field. We have now explained this approach in more detail in the text and have rewritten most of the noise section accordingly.
Response to Reviewer #2

We would like to thank the reviewer for his encouragement and positive assessment of our manuscript and the suggestions to improve the manuscript, which we have taken into account. Here we respond to all the comments made by the reviewer and indicate the changes in the manuscript made accordingly.

Several typographical errors to correct: 1) Overall, many missing commas. When starting a sentence with a prepositional phrase, separate it from the sentence with a comma. Page/Line 1/13 - For this reason, 7/11 - For the Lagrangian descriptor $M_v$, 7/16 - ... in a flow, 9/26 - For $M_v$, 14/4 - However, nowadays, 14/12 - However,

In some cases, there are commas that are unnecessary, page 5 line 8: "...dynamical evolution yield" (no comma).

Do a re-read of the paper and look for these prepositional phrases and clearly separate them grammatically.

We have done our best in re-reading the manuscript and improving grammar. Since we are not native English speakers, we admit having some problems with that.

At times, the papers language is too conversational. In general, the tone of the paper is scientific, and it should remain so throughout the paper. From above, "nowadays" is an example. Page/Line 3/15 - "Anyhow" can simply be removed. 7/6 - remove "Again".

Do a re-read and it helps to read it out loud so that you can catch the conversational tone when it comes up. That said, the language is outstanding for a non-English native speaker!

Thank you very much for pointing out the conversational language. We have removed it, whenever we noticed it ourselves.

Several words can be removed as they are unnecessary. In some cases, words need to be added or changed. Page/Line 1/14 - change "e.g. marine biology." to "marine biology for instance." 2/21 - "...but have been recently..." 4/31 - change on to of 5/10 - Change: "Manifold trajectories on both sides of the manifold have different behaviors compared..." 8/6 - change "distinction" to "distinguishing between ... and the identification " 9/2 - change "We use its" to "We use the feature" 14/3 - Change "non" to "none"

We have changed the text accordingly.

Check your formulae: Page/Line 5/3 - equation #3 - make sure the velocity is squared and the $dt$ is not under the $\sqrt$.

We have corrected the formula.
My main problem with this paper is that it asserts things that it does not support directly in the text. At times, there are conflicting statements about what the newly proposed Euler-Lagrangian descriptor can and cannot do. These discrepancies need to be resolved in the text so that the reader is not confused or led astray. Also, in the beginning of the paper, the use of oceanographic data is discussed, but the paper is essentially about toy-models. I understand the need to verify a new metric by using toy-models, however, if you suggest that this metric is useful for geo-physical flows, then you need to demonstrate that in this paper, or put a disclaimer early within the text, that you intend to follow-up this paper with another paper demonstrating the metric on actual physical flows obtained from either satellite data or well-understood simulated oceanographic models such as CCSM4 or a variant of ROMS that is well-accepted as a good representation of historical data (flows).

We agree completely with the reviewer that a demonstration of the method with a real oceanographic velocity field is much more convincing. To apply the Lagrangian descriptor $M_v$ based on the modulus of vorticity and the eddy tracking employing it to an oceanographic velocity field, was already planned when we submitted the manuscript. Now, we have included an example for the western Baltic Sea in the revised version and replaced the Section about the seeded eddy model. Furthermore, we discuss on the basis of this example advantages and disadvantages of the method if it is applied to an oceanographic data set. The velocity field for the western Baltic Sea is from the ocean model described in Gräwe et al. (2015a). Further research aims at an eddy statistics for lifetime, size and track of eddies for the central Baltic Sea with the eddy tracking based on $M_v$. But this work is beyond the scope of this current manuscript.

Finally, when you compare your new metric to existing metrics, then state they your metric is better, you need to clearly state the differences and exactly HOW your metric out performs another. That is simply now done well in the text as it currently stands.

We have rewritten and complemented the parts of the text where we explain what the new metric searches for and what are the differences to existing methods. We hope it is now easier to understand and more precise. We now discuss in more detail the problems arising when applying this metric to a real oceanographic field. This sheds more light on the difficulties of an automated eddy detection. The comparison of the results obtained with $M_v$ and the eddy tracking toolbox by Nencioli et al. (2010) reveals that for both methods false positives and false negatives exist. To improve those results is a future challenge.

From my understanding of $M_v$, you state its superiority over the $M$-value mainly because it maximizes when a fluid packet is part of gyre. In this case, for the duration of its stay within the gyre, the vorticity is high so the $M_v$ will be maximal. For the $M$-value, the center of the gyre will be a minimum, such that the $M_v$ can distinguish an elliptical point as well as a hyperbolic point, whereas, the $M$-value shows both types as minima. That is the main difference you quote in their behavior.
First, you state that your metric has excellent time resolution when seeking the beginning of a gyres formation as well as its lifetime, because you can measure when the gyre dies off. In both of these measurements, you depend on the value of tau. You make a cases in figure 6 that the best value for tau is 0.15 times the lifetime of the gyre. This is a circular definition. You need to know the lifetime in order to determine tau if it is to be based on a percentage of that lifetime. Furthermore, you can only find a gyre once you vorticity values are maximized, meaning that you need a particle to have already been inside of a gyre long enough for the M_v to become maximal. This means that there is a lead-in time where you do not know whether you are in a gyre or not as the trajectory has not had enough time to sample to gyre. The problem is that in order to find the gyre in the first place, you need an initial value for tau simply to compute the M_v. So, do you propose to constantly be computing M_v for a range of tau values until you find a gyre - THEN you can adjust tau to be 0.15 the lifetime of the gyre? But wait, you need to know the end of the gyre as well to know the lifetime, so you cannot determine an optimal tau to find a gyre until it has formed and gone away. This suggests that an oceanographer will need to be computing M_v over a range of tau values constantly simply to see when/if a gyre has formed. Of course, this is also true for the M-value.

The proper choice of tau is indeed the main problem with any Lagrangian descriptor including M_v and M. Hence, for a real oceanographic problem one has to vary tau to find all the eddies. This necessary choice is a practical limitation of the method. We point to that fact now better in the manuscript and provide an improved figure to show the dependence on tau (Fig. 6).

Figure 3 needs to be larger and with a better color contrast in order to show the manifold structure.

We have changed the colorcode to improve the contrast, because a larger tau does not lead to a clearer structure. Unfortunately, the colorcode does not take into account colour-blindness, but we did not find any colorcode with enough color-dimensions that is also valid for color-blindness.

Page 5, lines 6-15. At the beginning of the paragraph, you state that the M-value can distinguish between stable and unstable manifolds as well as hyperbolic and elliptic regions. On line 14, you state that the M-value cannot distinguish between elliptic and hyperbolic points.

Manifolds as well as hyperbolic and elliptic fixed points (more general distinguished hyperbolic trajectories and distinguished trajectories surrounded by an elliptic region in the sense of Mancho et al. (2013) correspond to singular features in the plot of M (singular lines and local minima). In this sense M can identify them. We apologize for the misleading use of the term “distinguish” on page 5 line 6 it was meant in the sense of “identify” (We are not native speakers.). We have rewritten the section on the Lagrangian descriptor M to make clear that M can of course identify distinguished hyperbolic trajectories and distinguished trajectory surrounded by an elliptic region but they are both displayed as a local minimum of M from which one cannot decide if it has elliptic or hyperbolic properties. Therefore, we constructed a
vorticity based Lagrangian descriptor $M_v$ that yields singular lines and local minima and maxima where the local maxima correspond to eddy cores (moving elliptic points) and the local minima to the distinguished hyperbolic trajectories (“moving saddle point”).

Page 7, figure 2. It is implied in the previous literature as well as your own figures, that the $M$-value is good at finding the radius of the elliptic regions because it has a minimum as the center, so that the contour of $M$-value maximizes as it moves away from the center and then decays as it moves far away - such that the maxima of $M$-values could be used to estimate the radii of elliptic regions. This is not explained in your paper, yet, you regularly refer to needing to use both the $M_v$ and $M$-value to extract useful gyre information. Page 9, lines 3 and 4 - refer to using the $M_v$ in combination with the $M$-value. Pages 15-17 also make it unclear in all of the figures which $M$ function is used to extract the gyre location AND SIZE. In the figures, is it stated $M$ and $M_v$. Why both?

As explained above $M$ and $M_v$ yield singular features (singular lines and local minima and maxima).

The eddy core in case of $M$ corresponds to a local minimum and in case of $M_v$ to a local maximum. Because the Lagrangian descriptor $M$ would display a minimum in case of a DHT too a second criterion is needed to distinguish them properly. Therefore, we suggest $M_v$ to simplify the automated eddy detection because one has only to search for a local maximum that corresponds to the eddy core.

The local maxima and the singular lines of $M_v$ will be used to construct an eddy tracking tool based on the following concept of an eddy: We denote an eddy as being bounded by pieces of stable and unstable manifolds of DHTs (according to Branicki et al. (2011) and Mendoza and Mancho (2012)) surrounding an area in which the flow is rotating. The manifolds correspond to singular lines in $M_v$ which are used to describe the eddy boundaries. The eddy core is considered as a local maximum of $M_v$ within this bounded region, which can be interpreted as one point of a distinguished trajectory surrounded by an elliptic region.

For the detection of the eddy shape we have previously used a combination of $M$ and $M_v$ because $M$ shows in our test case a clear line of minimum $M$ values that was easier to detect automated than the line in $M_v$. In general, manifolds correspond to singular lines (Mancho et. al. 2013). To construct an eddy shape detection that is more general and only based on $M_v$, we have improved the shape detection algorithm. The improved shape detection is based on the assumption that the eddy boundary is the largest closed contour line of $M_v$ where $M_v$ is an extremum (large gradient of $M_v$).

Furthermore, we have rewritten the Sections 2 and 3 to clarify the idea of $M$ and $M_v$ and its correspondence to our understanding of an eddy.

Page 8, lines 6-10. This paragraph asserts that $M_v$ is the best of four metric because it can
discern between stable and unstable manifold lines - which can be used to get more insight into the size of the eddies. HOW exactly? I feel like a paragraph explaining this statement is missing. Perhaps it would precede this paragraph.

We have rewritten this paragraph and parts of Sect. 2 to make clear what the idea of the description of an eddy boundary based on manifolds is, namely to describe a region that is separated from the rest of the flow (as explained above and in Branicki et al. (2011) and Mendoza and Mancho (2012)).

*Can M_v distinguish between stable and unstable manifold lines? If so, how?*

Singular lines in the plot of M or M_v correspond to manifolds. But one cannot distinguish based on the plot of M or M_v if it is a stable or unstable manifold. For the understanding of an eddy as a region bounded by pieces of stable and unstable manifolds of the distinguished hyperbolic trajectory (“moving saddle point”) with an eddy core inside, it is only necessary to identify the manifolds and not the type of the manifold. Furthermore, if one is interested in the type of the manifold one can put tracers on the manifold close to the hyperbolic trajectory and track them forward and backward in time.

*For that matter, in Figure 2, you show the four convective cell case, where M_v maximizes at the center. As tau increases, the maxima form a flatter and flatter plane centered on the gyre. Doesn’t this make you less sensitive to the size of the gyre, not more sensitive? How does M_v determine the radius of a gyre. I’d like to know based on the text provided.*

The maximum of M_v does not form a flatter and flatter plane in the centre, instead the centre becomes sharper and sharper as minimum of M in figure 2 f). This cannot be seen so clear in the colorcode used because the maximum is a light yellow point in a yellowish region. As mentioned above we have changed the colorcode to improve the contrast.

*Page 11, figure 6. The resolution shown for this figure does not convince me that 0.15*lifetime is the optimal tau value. It could be any value from 0.06 up to 0.21*lifetime. There should be many more points to determine the best value.*

We have improved the figure and calculated more values. The chosen value of tau=0.15*lifetime is in our case the beginning of a 0.02 small region of the optimal tau values. We have chosen the lower bound of this region to minimize the computational effort for calculating M_v.

*Finally, in the beginning, I thought I was going to see this metric applied to an oceanographic data set. By the end, I did not find it. Please show me something geo-physical or tell me that it is coming in a later publication.*

We have applied the method to an example of the western Baltic Sea to give an outlook on the application to oceanographic data sets.
Response to Reviewer #3

We would like to thank the reviewer for his/her thorough analysis of our paper and the suggestions given to improve the manuscript. We have addressed all the concerns of the reviewer and have rewritten the manuscript accordingly. In the following we give detailed answers to the questions raised and indicate the changes in the manuscript made as a response to the reviewers suggestions.

There is a huge literature on the problem of eddy detection, coming from very different scientific communities. Thus it is increasingly complicated to do something really new and to do justice to the vast literature. I should recognize, however, that the authors do a reasonable summarizing job in their introduction. Unavoidably, there are important recent results missing. From the part of the literature I know, I feel the following two references merit some citation and discussion: Karrasch D, Huhn F, Haller G. 2015 Automated detection of coherent Lagrangian vortices in two-dimensional unsteady flows. Proc.R.Soc.A 471: 20140639. http://dx.doi.org/10.1098/rspa.2014.0639 Haller G., Hadjighasem A, Farazmand M, Huhn F Defining Coherent Vortices Objectively from the Vorticity http://arxiv.org/abs/1506.04061

We would like to thank the reviewer for pointing out these two recent publications. The first of them we already cited in the first version of our manuscript. We were not aware of the second paper, since it had not appeared yet in a peer-reviewed journal. In a comment to our manuscript, this paper was also pointed out to us and we have now cited it. Additionally, reviewer #1 suggested several other citations, which we have included too.

There is a number of imprecise or even false statements in the paper. Here is a selection of them: * p. 2, lines 26-28: It is stated that algorithms to find DHT rely on 'Lagrangian descriptors'. Please note that DHTs were defined and computed many years before the introduction of the Lagrangian descriptors. * p. 2, line 31: This sentence makes no sense: 'The unstable manifolds are often called material lines in 2d () and surfaces in 3d flows ()' * In many places the authors use the word 'fixed point' for what are special elliptic or hyperbolic trajectories (moving, and then not fixed at all): abstract, pages 5, 6, 7, 8, 9, 15, 17 ... this is deeply misleading.

We agree with the reviewer, that the concept of DHT has been developed earlier and have changed the corresponding formulation in the text.

That the unstable manifolds are often called material lines is taken from the literature, where one can find these formulations rather often. But according to the suggestion of the reviewer we have removed this sentence since it does not add to the content of our manuscript.

We thank the reviewer for pointing out the misleading formulation of fixed points in the manuscript. We have rewritten the whole text for the introduction of the Lagrangian
descriptor to avoid any confusion. Indeed, when we were writing about fixed points we meant indeed the elliptic and hyperbolic trajectories. We have changed that in the current version of the manuscript.

In Mancho et al 2013 it is clearly stated that essentially any fluid property can be integrated along trajectories and provide a suitable ‘Lagrangian descriptor’. In this sense the use of the vorticity is just another example of ‘Lagrangian descriptor’. I find the name ‘Euler-Lagrangian descriptor’ and the emphasis given in the discussions to the mixed character rather inadequate.

We completely agree with the reviewer and also reviewer #1 that already Mancho et al. 2013 pointed out, that any fluid property can be used to construct a Lagrangian descriptor. This has been mentioned by us explicitly already in the first version of the manuscript (cf. the sentence “As already pointed out by Mancho et al. (2013) any intrinsic physical ...” in the beginning of the paragraph before formula (4)). Our motivation to introduce the name Euler-Lagranian descriptor was a more practical one. Because we found it difficult to read talking about one and another Lagrangian descriptor, we introduced a distinction by the names Euler-Lagrangian for one of them and Lagrangian for the other. Since this has been found misleading by two reviewers (#1 and #3) because it would look like the definition of a new descriptor, which indeed is not the case as we of course know, we have changed this in the revised version. To emphasize this we have now avoided the name Euler-Lagrangian descriptor in the title and throughout the whole manuscript. Furthermore we have even more than before emphasized that the original idea had been already formulated in Mancho et al. 2013. We now write “We would like to emphasize, that it has been already pointed out by Mancho et al. (2013)…”.

I hardly can see any ’manifold’ in the plots of M and specially of M_v in Fig. 3. Perhaps tau=0.15 is too small, or the contrast of the figure is not enough.

We have changed the colorcode to improve the contrast, because a larger tau does not lead to a clearer structure. Unfortunately, the colorcode does not take into account colour-blindness, but we did not find any colorcode with enough color-dimensions that is also valid for color-blindness.

At a first sight it looks incorrect to say that M, at variance with M_v, can not distinguish between elliptic and hyperbolic areas, since in any plot of M one can clearly identify them. But after some thinking I recognize that there is a real advantage (perhaps the only one) of M vs M_v, which is the fact that ellipticity and hyperbolicity are simply assessed by the maximum or minimum character of M_v, much more easy to automatize that the more complex neighbourhood exploration needed for the case of M. But then I do not understand (and the authors do not give any hint of it) why in Section 4 they say they need a combination of M_v and M, instead of just M_v.

As the reviewer pointed out, the possible distinction between elliptic and hyperbolic points (more general distinguished hyperbolic trajectories and distinguished trajectories surrounded by an elliptic region in the sense of Mancho et al. (2013)) is the most important property of
M_v to make the detection of eddy cores in flows easier, since one does not need an additional criterion to discern distinguished hyperbolic trajectories and distinguished trajectories surrounded by an elliptic region in the sense of Mancho et al. (2013) as if one would use M.

For the detection of the eddy shape we have used previously a combination of M and M_v because M shows in our test case a clear line of minimum M values that was easier to detect automated than the line in M_v. In general manifolds correspond to singular lines of M and M_v (Mancho et al. 2013). To construct an eddy shape detection that is more general and only based on M_v, we have improved the shape detection. The improved shape detection relies on the assumption that the eddy boundary is the largest closed contourline of M_v where M_v is an extremum (large gradient of M_v). We have now rewritten the text accordingly and use now only M_v for the detection of the shape.

*I think that the most original part of this research is the assessment of the behaviour of the different indicators under different types of noise. Nevertheless, the definitions of noise types in page 11 are all incomplete: for type 1 and 2 one can not reproduce the paper results unless the authors define ‘noise strength’, given that for white noise this would depend on the particular spatial and temporal discretizations used, which are not completely stated. For type 3, it is only after reading a comment in the Supplemental material that one begins to understand that noise is added to the functions h1 and h2, but again, ‘strength’ or ‘noise level’ should be properly defined.

We have rewritten the definition of the noise to clarify how the noise is applied and what we take as a noise strength.

*In the Supplemental material, Sect. S1 there is no indication on how the time-dependence, needed to define T_c, is introduced in the seeded eddy model. Also I find very convoluted (and not well explained) the way the radii of the eddies are sampled. Since at the end the authors restrict to 15-25 km radii, it seems to me that all this complexity is irrelevant and that anything uniform or Gaussian in that range will give the same results.

The eddies in the seeded eddy model live infinite (which is not very realistic). We have chosen this setup because we were only interested in the detection of the different eddy shapes. The way of choosing the eddy radii was due to the fact that we would like to use the same distribution of radii as Abraham (1998) but restrict ourselves to a specific part of the distribution. We agree that one can even use a simpler distribution, because the model is already very artificial and simplified. In the revised version with have changed that completely: We have removed this rather artificial example of the eddy seeded model and replaced it by an example of a velocity field of the western Baltic Sea according to the suggestions of reviewer #2.

*Errata: * There is a missing square of the velocity in Eq. (3) * Page 12, line 14: signal to
noise ratio small? or large?

Thank you very much for pointing to those typos. We have made the corresponding corrections.
Response to short comment #1, #2 and the editor comment:

Comment #1: We would like to thank George Haller for pointing out this manuscript to us, which we were indeed not aware of. We first got to know that he and his group are working on similar things as us on a workshop in Potsdam, were the first author of the manuscript (R.V.-K.) presented a poster about the content of this manuscript. The method suggested by Haller et al. 2016 is very similar to ours, which has been developed independently based on a different concept. There are several differences between the two methods. Each of them has still difficulties in finding all eddies, which are partly short lived. Haller et al. 2016 show with their examples that one can indeed nicely identify eddies in ocean flows, which are large and long lived. However, for many applications the real challenge is to identify and track small and short-lived eddies. Our method is a contribution in this direction and as such a step forward towards finding robust methods dealing with the census of eddies.

In the revised version we have included the analysis of an oceanographic flow in the western Baltic Sea. We computed all eddies including their shapes and compare our results to the ones obtained by an eddy tracking tool box by Nencioli et al. (2010). This example reveals the advantages and disadvantages of both methods.

Comment #2 and editors comment: The main problem, which has been addressed by the second comment and answered partly by the editor comment, is the problem of objectivity. This discussion is going on in the community for a long time and is reflected by the comments and the reviewer #1 suggestions. We would like to emphasize that objectivity is not the main focus of our manuscript and it is not our goal to resolve this long-lasting debate. Our goal is to provide oceanographers with a tool, which makes it possible to do a reasonable census of vortices in oceanographic flow fields. This can be provided by several methods and we have added another one which is in fact the same as proposed by Mancho et al. (2013) but using another quantity for the construction of the Lagrangian descriptor.

Objectivity is an important mathematical property as pointed out by George Haller, since it allows to define Lagrangian coherent structures independent of the frame. But in her answer Ana Mancho discusses, that one has of course to expect that a coordinate transformation would change M, but would still give the right answer in the coordinate system used (cf. frame invariance section in Mendoza and Mancho (2012)). Therefore, we think that both approaches are valid in the context of their framework and one has to see them in relation to the coordinate system in which they are computed. The same applies in principle to local Lyapunov exponents as another method to identify Lagrangian coherent structures. They are also not invariant with respect to coordinate transformations. After a coordinate transformation one would get other local or finite time Lyapunov exponents, which still give reasonable answers to the original problem. Only the long-term Lyapunov exponents computed as time goes to infinity are unique characteristics of a chaotic process.

From the application point of view, it is in many cases important, that a certain suitable coordinate system is used to analyze a problem in this particular coordinate system. This is for instance true for oceanographic problems which are given in the earth’s coordinates where e.g. coastlines are well defined boundary conditions. Talking to oceanographers it turns out that they find objectivity of being of secondary importance since the coordinate system of the earth is given and rotations of this coordinate system means e.g. rotations of coastlines, which might not be useful in many contexts. When identifying eddies in ocean flows one always has to deal with those boundary conditions since the computation of trajectories of particles contains the problem that particles reach those boundaries. One has to solve this problem by either reflecting them or losing them for the rest of the computation. This poses an additional problem which has not been addressed in many algorithms.

Response to short comment #1, #2 and the editor comment
The submitted manuscript does not claim to solve this problem of objectivity and this is not at all the aim of this submission. The aim is to step forward in providing suitable tools for the census of eddies, a problem, which is of increasing interest in oceanography. None of the tools we have checked including ours so far are good enough to solve those problems in oceanography, they all have their pro’s and con’s, they might be frame-dependent or not. The deficiencies become even more pronounced when looking at data, which are corrupted by large noise. Despite of that one still wants to get some reasonable results on the lifetime, the tracks and the shape of vortices. To solve this problem is a task for the whole community working on the identification of Lagrangian coherent structures.
Detecting and tracking eddies in oceanic flow fields: A vorticity Lagrangian descriptor based Euler-Lagrangian method on the modulus of vorticity

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Abstract. Since eddies play a major role in the dynamics of oceanic flows, it is of great interest to detect them and gain information about their tracks, their lifetimes and their shapes. We develop a vorticity based heuristic Euler-Lagrangian descriptor utilizing the idea of Lagrangian coherent structures present: a Lagrangian descriptor based on the modulus of vorticity to construct an eddy tracking tool. In our approach we define an eddy as a region around an elliptic fixed point (eddy core) surrounded by manifolds (rotating region in the flow possessing an eddy boundary core corresponding to a local maximum of the Lagrangian descriptor and enclosed by pieces of manifolds of distinguished hyperbolic trajectories (eddy boundary). We test the performance of an eddy tracking tool based on this Euler-Lagrangian-Lagrangian descriptor using an convection flow of four eddies, a synthetic vortex street and an eddy seeded model velocity field of the western Baltic Sea. The results for eddy lifetime and eddy shape are compared to the results obtained with the Okubo-Weiss parameter, the modulus of vorticity and an eddy tracking tool used in oceanography. We show that the Euler-Lagrangian-vorticity based Lagrangian descriptor estimates lifetimes closer to the analytical results than any other method. Furthermore we demonstrate that eddy tracking based on this descriptor is robust with respect to certain types of noise which makes it a suitable tool: method for eddy detection in velocity fields obtained from observation.

1 Introduction

Transport of particles and chemical substances mediated by hydrodynamic flows are important components in the dynamics of ocean and atmosphere. For this reason, there is an increasing interest in identifying particular structures in the flows such as eddies or transport barriers to understand their role in transport and mixing of the fluid as well as their impact on e.g. marine biology: marine biology for instance. Of particular interest in oceanography are eddies, which can be responsible for the confinement of plankton within them and hence, important for the development of plankton blooms (Abraham (1998); Martin et al. (2002); Sandulescu et al. (2007)). Such eddies possess a large variety of sizes and lifetimes. To tackle the problem of recognizing such eddies in aperiodic flows, different approaches have been developed: on the one hand, there are several methods available which are inspired by dynamical systems theory (Haller (2015); Haller (2015); Mancho et al. (2013) and references...
therein), on the other hand, numerical software for automated eddy detection has been developed in oceanography based on either physical quantities of the flow (Okubo (1970); Weiss (1991); Okubo (1970); Weiss (1991); Nencioli et al. (2010)) or geometric measures (Sadarjoen and Post (2000)).

Algorithms for finding eddies in fluid flows are applied in very different fields of science such as in atmospheric science (Koh and Legras (2002)), celestial mechanics (Gawlik et al. (2009)), biological oceanography (Bastine and Feudel (2010); Huhn et al. (2012)) and the dynamics of swimmers (Wilson et al. (2009)). The largest field of application is oceanography, since oceanic flows contain a large number of mesoscale eddies of size 100-200 km, which are important components of advective transport. Their emergence and lifetime influences the transport of pollutants (Mezić et al. (2010); Olascoaga and Haller (2012); Tang and Luna (2013)) or plankton blooms (Bracco et al. (2000); Sandulescu et al. (2007); Rossi et al. (2008); Hernández-Carrasco et al. (2014)). There is an increasing number of eddy resolving data sets available provided either by observations (Donlon et al. (2012)) or by numerical simulations (Thacker et al. (2004); Dong et al. (2009)). Subsequently there is a growing interest in the census of eddies, their size and lifetimes depending on the season. This task requires robust algorithms for the computation of eddy boundaries as well as the precise detection of their appearance and disappearance in time based on numerical velocity fields (Petersen et al. (2013); Wischgoll and Scheuermann (2001); Dong et al. (2014)) as well as altimetry data (Chaigneau et al. (2008); Chelton et al. (2011)). However, the huge amount of available data poses a challenge to data analysis. As pointed out in Chaigneau et al. (2008) mesoscale and submesoscale eddies cannot be extracted from a turbulent flow without a suitable definition and a competitive automatic identification algorithm. Several such algorithms have been developed based on the various concepts mentioned above. In the following we will briefly discuss several of those algorithms.

Based on dynamical systems theory, one can search for Lagrangian coherent structures (LCS) which describe the most repelling or attracting manifolds in a flow (Haller and Yuan (2000)). The time evolution of these invariant manifolds make up the Lagrangian skeleton for the transport of particles in fluid flows. LCS can be considered as the organizing centres of hydrodynamic flows. Their computation is based on the search for stationary curves of shear in case of hyperbolic or parabolic LCS. Elliptic LCS like eddies are computed as stationary curves of averaged strain (Haller and Beron-Vera (2013); Karrasch et al. (2015); Onu et al. (2015)) or Lagrangian-averaged vorticity deviation (Haller et al. (2016)). Other methods to determine whether an eddy can be identified in the flow employ average Lagrangian velocities (Mezić et al. (2010)) or burning invariant manifolds (Mitchell and Mahoney (2012)). The latter have been introduced originally to track fronts in reaction diffusion systems (Mahoney et al. (2012)) but have been recently extended to the detection of eddies (Mahoney and Mitchell (2015)). A completely different approach which connects geometric properties of a flow with probabilistic measures utilizes transfer operators to identify LCS (Froyland and Padberg (2009)). Another, more heuristic approach is the computation of distinguished hyperbolic trajectories (DHT) and their stable and unstable manifolds to identify Lagrangian coherent structures in a flow. DHTs can be considered as a generalization of stagnation points of saddle type and their separatrices to general time-dependent flows (Ide et al. (2002); Wiggins (2005); Mancho et al. (2006)). Algorithms to compute DHTs and their manifolds rely on the computation of the ridges of Lagrangian descriptors such as e.g. arc lengths along trajectories of particles in the flow can be computed using Lagrangian descriptors, which integrate intrinsic physical properties for a finite time and thereby reveal the geometric structures in phase space (Mancho et al. (2013)). Stable and unstable manifolds can also be calculated using the ridges of finite time or
finite size Lyapunov exponents (FTLE or FSLE) (Artale et al. (1997); Boffetta et al. (2001); d'Ovidio et al. (2004); Artale et al. (1997); Boffetta et al. (2001); d'Ovidio et al. (2004); Branicki and Wiggins (2010)) using the idea that initially nearby particles in a flow will move apart in stretching regions while they will move closer to each other in contracting regions. The unstable manifolds are often called material lines in 2d (Koh and Legras (2002)) and surfaces in 3d-flows (Haller (2001); Bettencourt et al. (2012); Freyland et al. (2012)), since particles would gather along these manifolds identifying them as barriers to transport. While most of the algorithms mentioned above possess the property of objectivity, i.e., they are invariant with respect to certain transformations of the Eulerian coordinate system, the method of Lagrangian descriptors lacks this property (Haller (2015)). Nevertheless, they are often used since the implementation of this method is easy and the computation is fast. This is particularly important when analysing large velocity fields with lots of LCS appearing and disappearing. Hence, the heuristic is appropriate to gain insight into oceanographic flows in a considerable amount of time. It has already been successfully applied to compute Lagrangian coherent structures in the Kuroshio current (Mendoza et al. (2010); Mendoza and Mancho (2010)) and in the Polar Vortex (de la Cámara et al. (2012)), in the North-Western Mediterranean Sea (Branicki et al. (2011)) as well as analysing the possible dispersion of debris from the Malaysian Airlines flight MH370 airplane in the Indian Ocean (García-Garrido et al. (2015)). In the recent years there has been some effort to derive Eulerian quantities which can be used to draw conclusions about Lagrangian transport phenomena (Sturman and Wiggins (2009); McIlhany et al. (2011); McIlhany and Wiggins (2012); McIlhany et al. (2015)). In oceanography, one of the most popular methods to identify eddies is based on the Okubo-Weiss parameter (Okubo (1970); Weiss (1991)). This method relies on the strain and vorticity of the velocity field, and has been applied to both, numerical ocean model output and satellite data (Isern-Fontanet et al. (2006); Chelton et al. (2011)). Often, the underlying velocity field is derived from altimetric data under the assumption of geostrophic theory. In this approach two limitations can appear. First, the derivation of the velocity field can induce noise in the strain and vorticity field. This is usually reduced by applying a smoothing algorithm, which might, in turn, remove physical information. Secondly, Douglass and Richman (2015) show that eddies can have a significant ageostrophic contribution. Thus, the detection might fail when relying on geostrophic theory. A slightly different approach was developed by Yang et al. (2001) and Fernandes et al. (2011), who used the signature of eddies in the sea surface temperature (SST) to detect them. Anyhow, the partially sparse coverage of satellite SST data limits the application of this method.

Sadarjoen and Post (2000) developed a tracking algorithm that is based on the flow geometry. The assumption is that eddies can be defined as features characterized by circular or spiral streamlines around the core of an eddy. The streamlines are derived from the velocity field. Additionally, the change of direction of the segments that compose the streamline (winding angle) is computed for each streamline. Chaigneau et al. (2008) applied this winding angle approach to a data set of the South Pacific. Moreover, they compared the winding angle method to the Okubo-Weiss approach and concluded that the former is more...
successful in detecting eddies and more important with a much smaller excess of detection errors. A further method based on geometric properties is proposed by Nencioli et al. (2010). The underlying idea is that within an eddy, the velocity field changes its direction in a unique way. Moreover, the relative velocity in the eddy core should vanish and should be enclosed by closed stream lines. This detection and tracking algorithm was successfully applied by Dong et al. (2012) in the Southern California Bight. In addition, the detection algorithm of Nencioli et al. (2010) has the advantage that its application is not limited to surface fields (Isern-Fontanet et al. (2006); Chelton et al. (2011); Fernandes et al. (2011)). Thus, it is possible to track eddies in the interior of the ocean, without any surface signature.

In this paper we develop an eddy detection and tracking tool based on the heuristic method of the Lagrangian descriptor introduced by Mancho and co-workers (Madrid and Mancho (2009); Mancho et al. (2013)). Instead of using the arc length of trajectories, For the purpose of automated eddy detection, we propose to use the modulus of the vorticity as the scalar quantity to be computed along a trajectory. We find this method combining the Eulerian and Lagrangian approaches to be more reliable in the detection of eddies and their lifetimes than other methods. More specifically we instead of using the arc length of trajectories. We compare our method to four others, namely the original Lagrangian descriptor using the arc length (Madrid and Mancho (2009); Mendoza et al. (2010)), an oceanographic method based on geometric properties of the flow field (Nencioli et al. (2010)) and detection tools which employ the Okubo-Weiss parameter (Okubo (1970); Weiss (1991)) and the vorticity itself.

The paper is organized as follows: Sect. [ ] briefly reviews the Eulerian concepts vorticity and Okubo-Weiss parameter, the Lagrangian descriptor descriptors $M$ based on the arc length and introduces our Euler-Lagrangian descriptor. To demonstrate $M_y$ based on the modulus of vorticity, To compare the performance of this method compared to the Lagrangian descriptor the two Lagrangian descriptors and the Eulerian concepts we use two simple velocity fields: the model of four counter rotating eddies and a modified van Karman-Kármán vortex street in Sect. [ ]. In Sect. [ ] we describe the implementation of the Euler-Lagrangian descriptor Lagrangian descriptor based on the modulus of vorticity as a tracking tool identifying eddy lifetimes (Sect. [4.1]) and compare the results again with the aforementioned other methods. In Sect. [4.2] we study the performance of the method in cases where we corroborate the velocity fields with noise to test the robustness of the method if applied to velocity fields obtained from observational data. Finally in Sect. [4.3] we compare the Eulerian and the Euler-Lagrangian Lagrangian view of the eddy shape with application to the modified van Karman-Kármán vortex street and to a velocity field which is similar to the one introduced by Abraham (1998) to study the impact of mesoscale hydrodynamic structures on plankton blooms from oceanography describing the dynamics of the Western Baltic Sea (Gräwe et al. (2015a)). We conclude the paper with a discussion in Sect[5]

2 From Eulerian and to Lagrangian methods to an Euler-Lagrangian method

The dynamics of a fluid can be characterized employing two different concepts: the Eulerian and the Lagrangian view. While the Eulerian view uses quantities describing different properties of the velocity field, the Lagrangian view provides quantities from the perspective of a moving fluid particle. Out of the variety of different Eulerian and Lagrangian methods mentioned
in the introduction, we recall here briefly only those concepts, which will be important for our development of a measure to identify eddies in a flow.

An Eulerian method to describe the circulation density of a velocity field in hydrodynamics is vorticity \( \mathbf{W}(x, t) \) defined as the curl of the velocity field \( v(x, t) \). The vorticity associates a vector to each point in the fluid representing the local axis of rotation of a fluid particle. It displays areas with a large circulation density like eddies as regions of large vorticity and eddy cores as local maxima.

Another Eulerian quantity is the Okubo-Weiss parameter \( OW \). It weights the strain properties of the flow against the vorticity properties and distinguishes so strain dominated areas from vorticity dominated one. The Okubo-Weiss parameter is defined as

\[
OW = s_n^2 + s_s^2 - \omega^2, \tag{1}
\]

where the normal strain component \( s_n \), the shear strain component \( s_s \) and the relative vorticity \( \omega \) of a two dimensional velocity field \( v = (u, v) \) are defined as

\[
s_n = \frac{\partial u}{\partial x} - \frac{\partial v}{\partial y}, \quad s_s = \frac{\partial v}{\partial x} + \frac{\partial u}{\partial y} \quad \text{and} \quad \omega = \frac{\partial v}{\partial x} - \frac{\partial u}{\partial y}. \tag{2}
\]

Eddies are areas having a negative Okubo-Weiss parameter with a local minimum at the eddy core because here the vorticity component outweighs the strain component, while strain dominated areas are characterized by a positive Okubo-Weiss parameter.

A Lagrangian view on the dynamics of the velocity field is given by the Lagrangian descriptor developed by Mancho and coworkers (Madrid and Mancho (2009)). A more general definition of the Lagrangian descriptor is outlined in Mancho et al. (2013). Here we focus on the Lagrangian descriptor based on the arc length of a trajectory. It is defined as

\[
M(x^*, t^*)_{\nu, \tau} = \left( \frac{1}{2} \sum_{i=1}^{n} \left( \frac{dx_i(t)}{dt} \right)^2 \right)^{1/2} \int_{t^*-\tau}^{t^*+\tau} dt
\]

with \( x(t) = (x_1(t), x_2(t), \ldots, x_n(t)) \) being the trajectory of a fluid particle in the velocity field \( v \) that is defined in the time interval \( [t^*-\tau, t^*+\tau] \) and going through the point \( x^* \) at time \( t^* \).

The Lagrangian descriptor \( M \) can distinguish stable and unstable manifolds as well as hyperbolic and elliptic regions in the velocity field at the same time yields singular features that can be interpreted as time-dependent “phase space structures” like (time-dependent or moving) elliptic or hyperbolic “fixed” points (denoted as distinguished elliptic or hyperbolic trajectories DET and DHT respectively in Madrid and Mancho (2009)) and their time dependent stable and unstable manifolds (Mancho et al. (2013), Wiggins and Mancho (2014)). The reason is that \( M \) accumulates different values of the arc length depending on the dynamics in the region. Trajectories that have a similar dynamical evolution \( \tau \) yield similar values of \( M \). When the dynamics changes abruptly, \( M \) will change too. This is the case at fixed points and distinguished hyperbolic trajectories (DHTs) and their stable and unstable manifolds. In case of manifolds trajectories on both sides of the manifold have a different behaviour compared to the behaviour of the trajectories on the
manifold. Either they approach the manifold very fast or the they move away from the manifold very fast. In both cases they accumulate a larger values of $M$ in a given time interval than trajectories on the manifold. Therefore the manifold is displayed as a. the singular line of $M$ in a contour color-coded plot of $M$. Fixed points themselves correspond to local minima can be interpreted as corresponding to a manifold. If a trajectory stays in a region or at one point it accumulates a small or zero value of $M$. Here the trajectories do not leave the region of $M$ becomes a local minimum. While DHTs have been extensively studied, distinguished trajectories possessing an elliptic type are less understood. However, such trajectories can also be identified as singular features of $M$ being surrounded by an elliptic region in the sense of [Mancho et al. (2013)]. For an extensive discussion about the notion of hyperbolic and elliptic regions in flows we refer to [Mancho et al. (2013)]. For each instant of time $t^*$ the fixed point and accumulate small values of color-coded plots of $M$. As a consequence the Lagrangian descriptor can be interpreted as a “snapshot” of the phase space, where the minima correspond to one point of a DHT or a distinguished trajectory surrounded by an elliptic region. When $t^*$ is changing $M$ cannot distinguish between elliptic and hyperbolic fixed points and is therefore not suitable to identify eddies without using a second criterion. Therefore the Eulerian and the Lagrangian view on the dynamics in a flow will be combined into a new quantity: reveals the time evolution of the phase space and loosely speaking distinguished hyperbolic trajectories can be considered as “moving saddle points”. distinguished trajectories surrounded by an elliptic region in the sense of [Mancho et al. (2013)] as “moving elliptic points”. Due to the arbitrary time-dependence of the flow, both, the DHTs and the distinguished trajectories surrounded by an elliptic region are time-dependent and exist in general only for a finite time in a time-dependent flow. Hyperbolicity in case of DHT refers to the fact that along those trajectories Lyapunov exponents are positive or negative, but not zero except for the direction along the trajectory [Mancho et al. (2013)].

As already Because the Lagrangian descriptor $M$ would display minima in both cases, i.e. DHT and distinguished trajectories surrounded by an elliptic region, a second criterion is needed to distinguish them properly. To avoid such additional distinction criterion, we suggest a Lagrangian descriptor based on the modulus of vorticity to simplify the automated eddy detection. We emphasize, that it has already been pointed out by [Mancho et al. (2013)] that any intrinsic physical or geometrical property of trajectories can be used to construct a Lagrangian descriptor by integrating this property along trajectories over a certain time interval. Therefore, we introduce a vorticity based Lagrangian descriptor $M_V$ in which the physical quantity is the modulus of the vorticity $W$ of a velocity field $v(x, t)$

$$W(x, t) = |\nabla \times v(x, t)|.$$ 

We define the Euler-Lagrangian Lagrangian descriptor $M_V$ as based on the modulus of vorticity as

$$M_V(x^*, t^*)_{\tau} = \int_{t^* - \tau}^{t^* + \tau} (W(x, t))^\gamma dt.$$ 

with $\gamma = 1/2$. The Euler-Lagrangian The Lagrangian descriptor $M_V$ combines the Eulerian and the Lagrangian view of a dynamical system by measuring based on the modulus of vorticity measures the Eulerian quantity modulus of vorticity along a trajectory (Lagrangian view) passing through a position $x^*$ at time $t^*$ in a time interval $[t^* - \tau, t^* + \tau]$. Within this time
The method \( M \) is chosen large enough. The choice of \( \tau \) depends on the parameter \( \tau \) that gives the length of the time interval. Structures that live shorter than \( 2\tau \) cannot be resolved.

Even structures that live longer than \( 2\tau \) can only be resolved if \( \tau \) is chosen large enough. The choice of \( \tau \) depends on the structure and the time scale of the flow field considered. Within the range of the time scale of the problem that should be resolved some variation of \( \tau \) is needed until the optimal \( \tau \) for a given problem is found.

3 Eddies in a flow: Comparing the Eulerian, Lagrangian and Euler–Lagrangian method

To compare the performance of the proposed Euler–Lagrangian method to others, two test cases, namely a convection flow consisting of four counter rotating eddies and a model of a vortex street are used. The four counter rotating eddies are used to show that different methods detect different aspects of the eddies. Additionally, we discuss how the displayed structure depends on the chosen \( \tau \). The vortex street is particularly used to test how suitable our method is to detect and track eddies in comparison to other methods and how well they all estimate eddy lifetimes and shapes. This way we gain insight into performance, advantages and disadvantages of the proposed method compared to the others.

To give a complete view of the advantages and disadvantages the results of the different test cases are interpreted in a coherent discussion after presenting all results.
The equations of motion of fluid particles in a convection flow of four counter rotating eddies are given by
\[ u = \dot{x} = \sin(2\pi \cdot x) \cdot \cos(2\pi \cdot y) \quad \text{and} \quad v = \dot{y} = -\cos(2\pi \cdot x) \cdot \sin(2\pi \cdot y). \] (6)

We compute the four different quantities, modulus of vorticity, Okubo-Weiss parameter, Euler-Lagrangian descriptor and Euler-Lagrangian descriptor, the two Lagrangian descriptors \( M \) and \( M_V \), on a spatial domain \( (0,1) \times (0,1) \times [0,1] \times [0,1] \). To this end, the spatial domain is decomposed into a discrete grid \( (201 \times 201) \) and the different methods are calculated for each grid point. The results are presented in Figs. 1 and 2.

The model of the vortex street consists of two eddies that emerge at two given positions in space, travel a distance \( L \) in positive \( x \)-direction and fade out. The two eddies are counter rotating. They emerge and die out periodically with a time shift of half a period. The model is adapted from Jung et al. (1993) and Sandulescu et al. (2006) with the difference that the cylinder as the cause of eddy formation and its impact on the flow field due to its shade is neglected. In this sense, the eddies emerge non-physically out of the blue nowhere, but all quantities like lifetime and radius to be estimated by means of eddy tracking are then given analytically and make up a perfect test scenario. A detailed description of the model can be found in the supplemental material to this article. Again all methods are applied to this velocity field using a \( (302 \times 122) \) grid. Unless otherwise stated, the time interval \( \tau \) for the Lagrangian and Euler-Lagrangian methods is set to 0.15 times the lifetime of an eddy. The results are presented in Fig. 3.

These two test cases reveal the following characteristics of the properties of coherent structures in a flow: Eulerian-Lagrangian methods are capable to identify eddy cores (moving elliptic fixed points). Euler-Lagrangian methods display eddy cores as local maxima (modulus of vorticity, \( M_V \)) or local minima (Okubo-Weiss, \( M \)) of the respective quantity (Fig. 1). The Lagrangian and Euler Lagrangian methods identify moving hyperbolic fixed points as local minima. Local minima of the Lagrangian methods correspond to DHTs (Fig. 2c-f). For the Lagrangian descriptor \( M \) the centre/core of the eddy as well as the moving hyperbolic fixed point-DHT are indistinguishable since they are both displayed as local minima of \( M \). Our Euler-Lagrangian-Lagrangian descriptor \( M_V \) based on modulus of vorticity can clearly distinguish between the centres/core of an eddy and moving hyperbolic points a DHT (Fig. 2e-f(a)-(c)). For this reason, Eulerian and Euler-Lagrangian methods are more appropriate than the Lagrangian descriptor \( M \) for identifying eddies in a general time dependent velocity field. An automated identification of eddies, since no further criteria are needed.

To characterize Lagrangian coherent structures in a flow not only elliptic fixed points distinguished trajectories surrounded by an elliptic region in the sense of Mancho et al. (2013) associated with eddy cores and moving hyperbolic fixed points DHTs have to be identified but also the stable and unstable manifolds associated with the latter to find eddy boundaries: according to the concept of an eddy in Sect. 2. Those manifolds are only identifiable visible as singular lines in the contour plot using color-coded plot of the Lagrangian descriptor \( M \) (Fig. 2d-f, 3c) and the Euler-Lagrangian-(d)-(f), 3(c) and the Lagrangian descriptor \( M_V \) (Fig. 2a-e, 3a-d(a)-(c), 3(d)) respectively.

How detailed the displayed fine structure of the Lagrangian descriptor descriptors \( M \) and the Euler-Lagrangian descriptor \( M_V \) is represented depends on the chosen value of the time interval \( \tau \). It ranges from no clear structure for small \( \tau \) (Fig. 2a and d(a) and (d)) to a detailed structure for large \( \tau \) (Fig. 2e and f(c) and (f)).
From these properties, distinction between elliptic and hyperbolic fixed points and identification of manifolds, we can conclude that the Euler-Lagrangian-Lagrangian descriptor $M_V$ is the most suitable quantity to compute when studying oceanographic flows with respect to a suitable method for an automated search for eddies. It is the only quantity in oceanographic flows. Out of the four quantities considered which considered quantities $M_V$ allows for a clear identification of eddy cores and the stable and unstable manifolds of hyperbolic fixed points which can be used DHTs necessary to get more insight into the size of eddies with the least number of check criteria. For this reason we suggest to use $M_V$ as the basis for an eddy tracking tool. How these properties of $M_V$ are implemented into an eddy tracking tool is explained for the eddy core in Sect. 4.1 and the eddy size and shape in Sect. 4.3.

4 Euler-Lagrangian-Lagrangian descriptor $M_V$ as eddy tracking tool

The mean oceanic flow is superimposed by many eddies of different sizes which emerge at some time instant, persist for some time interval and disappear. Consequently, an eddy tracking tool has to detect them at the instance of emergence, track them over their lifetime and detect their disappearance. To classify the different eddies some information about their size is needed too. This way one can finally obtain the time evolution of a size distribution function of eddies. In this section we apply the Euler-Lagrangian modulus of vorticity based Lagrangian descriptor $M_V$ to the hydrodynamical hydrodynamic model of a vortex street to test its performance as an eddy tracking tool. We use its characteristics that it displays the centres of the eddies (elliptic fixed points) as local maxima and allows simultaneously the identification of invariant manifolds in the flow which in combination with the Lagrangian descriptor $M$ will serve as estimators local maxima of $M_V$ for an automated search for eddy cores and in addition, the area enclosed by the singular lines of $M_V$ associated with the manifolds of the DHTs as measure for the size of the eddies.

4.1 Eddy birth and lifetime

We ask First we check how well $M_V$ predicts the lifetime detects the birth of an eddy and the time instant of the eddy birth its lifetime and compare the results to the oceanographic eddy tracking tool box (ETTB) by Nencioli et al. (2010), as well as Eulerian quantities like the Okubo-Weiss parameter and the modulus of vorticity. As pointed out in the previous section the Lagrangian descriptor $M$ cannot be used as an eddy tracking tool because it does not distinguish between elliptic and hyperbolic fixed points. However, it will be used to determine the size of the eddy, once it has been detected using the Euler-Lagrangian descriptor $M_V$.

The idea of the tracking inspired by Nencioli et al. (2010) is to search for local maxima ($M_V$ and modulus of vorticity) or local minima (Okubo-Weiss, velocity based method by Nencioli et al. (2010)) surrounded by a region of gradient towards the maximum or minimum in a given search window that is shifted to the region of interest. The size of the search window determines which maximal eddy sizes can be detected. The eddy is tracked from one time step to the other next by searching for an eddy core with the same direction of rotation within a given distance. The choice of this distance depends on the velocity field. It should be in the range of the maximal distance a particle could travel in the timespan of interest. The
position of an eddy is logged in a track-list for each eddy at each time step. To give a criterion when an eddy track is completed or to avoid false positive tracks or tracks of short living eddies one is not interested in, a threshold number of time steps should be defined. A track-list of an eddy that is not longer updated for this number of time steps is closed. A track-list that is shorter than the that is shorter than a given threshold number of time steps is deleted to focus on eddies which exist longer than this minimum time interval. A detailed description of the algorithm algorithms can be found in the supplemental material to this article.

In order to check the accuracy of the eddy tracking algorithm, we use the dimensionless model of the vortex street presented in Sect. 3 in a dimensionless parametrization, since the time instant of birth of the eddies and their lifetimes are given analytically. We measured both quantities for different dimensionless lifetimes $T_c$ and dimensionless vortex strengths of 200, 100 and 50 for the eddy that arises at time $T_c/2$. The rationale behind varying the vortex strength is to estimate how weak an eddy could be to have still a reliable detection by the method be still reliably detected by the method. For $M_V$, $\tau$ was chosen as $0.15 \cdot T_c$. The results are presented in Figs. 4 and 5.

In all cases independent of the vortex strength, the results obtained with $M_V$ are close to the analytical $T_c$ (Fig. 4) or the analytical time instant of birth (Fig. 5). All other methods underestimate $T_c$ and overestimate the time instant of birth. Especially in case of the eddy tracking tool box (ETTB) by Nencioli et al. (2010) ETTB the estimated times depend heavily on the vortex strength. For that method it becomes more and more difficult to detect the eddy as the rotation speed decreases. The reason for the good estimates provided by $M_V$ is that by construction $M_V$ lies in its construction which makes use of the history of the eddy (past and future). Hence it can detect eddies earlier than they arise by taking into account the future or detect them longer than they actually exist by looking into the past. $M_V$ is not restricted to the information about the velocity field at one instant of time like the other methods. However, the performance of $M_V$ depends on the chosen value of $\tau$ (Fig. 6). If $\tau$ gets too large in relation to $T_c$, the estimate of the lifetime deviates from the analytical one because the trajectories contain too much of the history of the eddy. There exists an a small range of optimal $\tau$ for a certain class of eddies. A good choice in our case is the range is between about 15 % and 18 % of the eddy lifetime. Other applications might need a larger or smaller $\tau$ or a $\tau$ that is a compromise between structures with very different lifetimes. It is also advisable to vary $\tau$ to detect different size and lifetime spectra of eddies.

4.2 Robustness of the lifetime detection with respect to noise

Velocity fields describing ocean flows have either a finite resolution when obtained by simulations or contain measurement noise when retrieved from observational data. For this reason, an eddy tracking method has to be robust with respect to fluctuations of the velocity field. Therefore, we explore how the detected eddy lifetime depends on noise in the velocity data.

We use three types of noise based on the different sources of noise. To test the influence of noise in a manageable test setup where we know all parameters like e.g. eddy lifetime (here $T_c = 1$) or vortex strength (here $w = 200$) we use the velocity components $u(x, y, t)$ and $v(x, y, t)$ of the vortex street mentioned in Sect. 3 and add three different types of noise...
to it mimicking measurement noise that can arise in observations or simulations. The result are noisy velocity components \( v_N(x, y, t) \) and \( v_N(x, y, t) \) for which we calculate Okubo-Weiss, modulus of vorticity and \( M_V \) and then apply the different eddy tracking methods. The noise is realised as white Gaussian noise in form of a matrix of normally distributed random numbers of the grid size for each time step multiplied by a factor that is referred to as noise level or noise strength. The noise level is given dimensionless, because the noise is applied to the dimensionless model of the vortex street presented in Sect. 3. The different noise types and their motivation are given in the following:

1. type 1: We add white Gaussian noise \( \xi(x, y, t) \) of different noise strength \( \sigma \) between 0.05 and 0.95 to the velocity field components \( u(x, y, t) \) and \( v(x, y, t) \) of the vortex street with analytical eddy lifetime \( T_e = 1 \) and vortex strength \( w = 200 \). The noise is uncorrelated in space and time. The velocity field resulting velocity components \( u_N(x, y, t) = u(x, y, t) + \sigma \cdot \xi_u(x, y, t) \) and \( v_N(x, y, t) = v(x, y, t) + \sigma \cdot \xi_v(x, y, t) \) in this case are still periodic but noisy. This type of noise mimics the effect of computing derivatives of observed velocity fields (e.g. by satellites or HF-radar).

2. type 2: We add multiplicative white Gaussian noise \( \xi(x, y, t) \) of different noise strength \( \sigma \) between 0.05 and 0.95 to the velocity field components \( u(x, y, t) \) and \( v(x, y, t) \) of the vortex street but take into account that the actual noise depends on the velocity itself by taking the maximum of it over the whole spatial grid. The motivation is that the strength of noise depends on the “Signal to noise” ratio. If we have a strong current, it is easy to detect this by a satellite, since the signal-strength is high. This is opposite for slow currents, where the noise level is much higher. Thus, we add white noise that is inversely proportional to the current speed in the sense of noise/(1 + maximum of the current speed). The noisy velocity components are given as \( u_N(x, y, t) = u(x, y, t) + \sigma \cdot \xi_u(x, y, t)/(1 + \max_{x,y}(u(x, y, t))) \) and \( v_N(x, y, t) = v(x, y, t) + \sigma \cdot \xi_v(x, y, t)/(1 + \max_{x,y}(v(x, y, t))) \).

3. type 3: We add white Gaussian noise \( \xi(t) \) of different noise strength \( \sigma \) between 0.05 and 0.5 to the \( y \)-component of the eddy centres’ movement. The equations of the unperturbed velocity field contains a part that describes the movement of the eddy centres (see supplemental material). The motion of the \( y \)-components of the eddy centres in the unperturbed velocity field \( (u, v) \) is given by \( y_1(t) = y_0 + y_2(t) \) where the index \( 2 \) refers to the two eddies. The movement of the eddy centres in case of noise is given by \( y_1(N(t)) = y_0 + \sigma \cdot \xi(t) \) and \( y_2(N(t)) = y_0 + \sigma \cdot \xi(t) \). This type of noise can be observed if the velocity fields have to rely on georeferencing. For instance, satellite generated velocity fields have to be mapped on a longitude/latitude grid, since the satellite is moving. During this postprocessing step a shift in the georeference is possible, leading to translational shifts and thus to type 3 noise. However, a high noise level of type 3 is not very likely. If one deals with typical geophysical applications, which have a grid resolution of the order 1 to 10 km, the georeferencing errors are mostly small compared to the grid cell size. We have applied noise of type 1, 2 and for this reason, the considered noise levels for type 3 to the dimensionless model of the vortex street presented in Sect. 3. We have used noise are smaller than for type 1 and 2.

To explore the impact of noise systematically, we have used different noise strengths. For each noise strength \( \sigma \) 1000 realizations of the white Gaussian noise per noise strength and were calculated. In the resulting velocity fields we estimated the
lifetime of each eddy that undergoes a whole life-cycle within the time of simulation. The plotted eddy lifetimes obtained with all different tracking methods are medians of the distributions of the lifetimes for the 1000 realizations per noise strength (Fig. 7, 8, 9).

The three types of noise illustrate different advantages and disadvantages of \(M_V\) compared to the other methods. In case of type 1 noise, \(M_V\) gives the best estimate of the lifetime compared to the all other methods independent of the increasing noise level. The reason why the error of the estimate in case of \(M_V\) does not increase with increasing noise level is that \(M_V\) is a measure that is based on an integral. It accumulates the Integrating over accumulated uncorrelated noise along the trajectory from past to future. Integrating over this accumulated noise can be considered as a smoothing process. Also the method ETTB by Nencioli et al. (2010) gives a good result independent of the increasing noise level, because the signal to noise ratio is small. The minimum of the velocity that is the key-signal for determining the eddy core in their method still remains a local minimum in the contour plot of the velocity. However, with increasing noise level we see an increase of outliers for the method ETTB by Nencioli et al. (2010) and \(M_V\) (boxplot not shown here) of the data of the method. The performance of modulus of vorticity and the Okubo-Weiss parameter decreases as expected with increasing noise level while the distribution increases in width (Fig. 7). The reason is that the noise gets so large that it increasingly disturbs the key-signal for an eddy core until no distinct eddy core can be identified anymore.

In case of type 2 noise, \(M_V\) and the method by Nencioli et al. (2010) ETTB show a similar behaviour as in case of type 1 noise. Both yield good results independent of the noise level. This is again due to the smoothing process in case of \(M_V\). The modulus of vorticity gives results that are even better than \(M_V\) in case of small noise levels, but its performance drops below the results of \(M_V\) with increasing noise level (Fig. 8). The reason is that the key-signal for determining an eddy core using the modulus vorticity is stronger in case of small noise levels and gets disturbed by the noise with increasing noise level. By contrast, the smoothing process in \(M_V\) is the same independent of the noise level. As expected, the performance of Okubo-Weiss decreases with increasing noise level. In contrast to type 1 noise, Okubo-Weiss can identify eddy cores even in case of strong noise. Eddy cores can be identified because the key-signal for an eddy core is less disturbed.

In case of type 3 noise, \(M_V\) yields an estimate of the lifetime with the largest error (Fig. 9). The reason is that in this case noisy trajectories that start close to each other diverge fast, while the ones with no noise have a similar dynamical evolution. This divergence due to noise leads to a loss of structure in space which that can be interpreted as a weakening of the correlation between neighbouring trajectories. This effect is strongest in case of \(M_V\) because it integrates over time and so neighbouring trajectories that have similar values of \(M_V\) in case of no noise yield very different values of \(M_V\) due to the divergence of the trajectories. As a consequence no clear structure in \(M_V\) can be identified. This effect increases with the noise level.

Also for the other methods noise of type 3 affects strongly the identification of the eddy core because the weakening of the correlation between neighbouring points disturbs the key-signal of an eddy core (a local minimum or maximum in a certain domain). The error in estimating the lifetime increases with increasing noise level. In all cases the number of outliers in the boxplot (not shown here) increases with the noise level.

As a result, non consequence, none of the methods performs in an optimal way when the noise affects the movement of the eddy cores. This disadvantage will lead to deviations in the lifetime statistics for
eddy tracking based on observational data. However, nowadays, the error in georeferencing of satellite images (which is mimicked by type 3 noise) is mostly small. For special applications, a georeferencing error of smaller than 1/50 pixel is achievable (Leprince et al., 2007). Eugenio and Marqués (2003) show that with reasonable effort a mapping error smaller than 0.5 pixel is possible, if fixed landmarks (coastlines, islands) are on the images. With the increase in earth orbiting satellites and thus the increase in available images, it can be assumed that this error will drop even more (Morrow and Le Traon, 2012). Anyhow, if numerically generated velocity fields are used, noise of type 3 is completely absent, if numerically generated velocity fields are used. Here the evolution of neighbouring trajectories are is smooth and correlated.

In summary, $M_V$ can be used for the detection of eddies and the estimate of eddy lifetimes for velocity fields with and without noise and yields good results independent of the noise level in case of type 1 and 2 noise. However, one has to take into account that the velocity field should not be too noisy and that one has to chose a $\tau$ that fits the problem. The Euler-Lagrangian Lagrangian descriptor $M_V$ has an additional advantage in detecting arising eddies earlier than other methods due to collecting information along the trajectory from past to future. This can be useful in the identification of regions that will be eddy dominated in the further evolution of the flow.

### 4.3 Detecting eddy sizes and shapes

Beside its lifetime an eddy is characterized by its shape. In the following we will estimate the eddy shape using a combination of the vorticity-based Euler Lagrangian size and shape using the Lagrangian descriptor $M_V$ and the arc length based Lagrangian approach $M$. We based on the modulus of vorticity and compare the results to the shape detected by the eddy tracking tool ETTB by Nencioli et al. (2010). In this way, we demonstrate the differences between the Eulerian and Euler-Lagrangian Lagrangian point of view on the eddy size and shape.

From the Euler-Lagrangian point of view the As mentioned in Sect. 2 the estimation of the eddy shape and size from the Lagrangian point of view is based on the idea that the boundaries of the eddy are linked to manifolds of DHTs that surround the eddy. This idea has been formulated by Bettencourt et al. (2012), who point out, that eddies are bounded by invariant manifolds surrounding the eddy and acting as transport barriers (Branicki et al., 2011; Bettencourt et al., 2012). These manifolds cannot be crossed by any trajectories and, therefore, trajectories starting inside the manifolds are trapped in the eddy.

Defining the boundaries in this way one can estimate the trapping region or volume that is transported by an eddy.

The shape detection algorithm combines the Lagrangian descriptor $M$ and the Euler Lagrangian Lagrangian descriptor $M_V$ and searches for displays singular lines that correspond to manifolds. Therefore, the shape detection algorithm searches for the largest closed contour lines with the local lowest level of $M$ which surrounds the elliptic fixed point of $M_V$ for which $M_V$ is an extremum and which surrounds an eddy core found with $M_V$. This line is contour line, extracted from $M_V$ with the MATLAB function contour and along which the gradient of $M_V$ is large, should be a line on or close to the manifold displayed by $M$ or very close to a singular line displayed by $M_V$. By contrast, the eddy tracking tool corresponding to a manifold and will give an estimate of the eddy boundary.

The ETTB by Nencioli et al. (2010) gives an Eulerian view on the eddy shape by defining the eddy boundaries as the largest closed streamline of the streamfunction around the eddy centre where the velocity still increases radially from the centre. The
contour lines as well as the streamlines are extracted in a given search window which is centred on the eddy core.
The comparison of the different views on the eddy size and shape is presented in Fig. 10 for the vortex street without (a) and with noise of type 1, 2 and 3 (b-d). As expected the shape (b)-(d)). The size detected with the eddy tracking tool ETTB by [Nencioli et al. (2010)] is much smaller than the shape size based on the Euler-Lagrangian-Lagrangian view (Fig. 10a-c). Additionally (a)-(c). Additionally, the evolution of the eddy is captured by both methods even in case of strong type 1 and 2 noise (Fig. 10b and e). The shape computed by the combination of $M$ and (b) and (c)). Here, the eddy boundaries in case of noise show small irregularities due to the noise. In general, the eddy boundary computed based on $M_V$ is detected earlier and shows more growing and shrinking during the evolution of the eddy than the eddy boundary extracted by the ETTB. This is due to the conceptual idea of the measure $M_V$ that contains the history of the trajectories. As shown in Sect. 4.2 this concept of $M$ and $M_V$ leads to problems in case of a velocity field with type 3 noise (although significant type 3 noise levels are very unlikely). If the noise level is too large, a structure, neither a clear eddy core nor a clear eddy boundary can be detected (Fig. 10d) with $M$ and (d) within $M_V$. But if an eddy core can be detected as in case of the left eddy in Fig. 10(d) the eddy shape detection based on $M_V$ gives an idea of the size and the noisy eddy boundary.

In a real oceanic flow eddies of different lifetime, size and shape will occur simultaneously. To illustrate, as an outlook, how different eddy shapes and sizes can be detected in real oceanic flow fields, we apply our approach to a seeded eddy model that mimics such an oceanic eddy field. The model is inspired by velocity fields of the western Baltic Sea for May 2009. The Baltic Sea is a good testbed, since the tides are negligible and the entire eddy dynamics is driven by baroclinic instabilities, frontal dynamics and the interaction with topography. An extended eddy statistics in the central Baltic Sea based on $M_V$ will be the content of further research.

A triple-nested circulation model was used to simulate the flow fields in the seeded eddy model of a turbulent flow described in [Abraham (1998)] and the model of [Jung et al. (1993)] to elucidate the role of eddies in the formation of plankton patchiness. The model consists of a background flow of 0.18 m/s and ten mesoscale eddies of different sizes. Their radii $r$ are drawn from a distribution possessing a power law $1/r^3$ where the minimal radius is 10 km and the maximal radius is western Baltic Sea. The innermost model domain was discretised in the horizontal with a spatial resolution of 1/3 nautical mile (approx. 600 m).

The model domain covers the Danish Straits and the western Baltic. The open boundaries are located in the Kattegat and at the eastern rim of the Bornholm Basin. In the vertical, 50 km. Because we are interested in mesoscale eddies, the drawn radii are between terrain-following adaptive layers, with a zooming toward stratification layers were used. The setup is identical to the one used by [Klingbeil et al. (2014)] or [Gräwe et al. (2015)]. There, a detailed description and validation of the present setup case be found. At the open boundaries of the model domain, the water elevations, depth-averaged currents, as well as salinity and temperature profiles are prescribed. This external forcing was taken from a model of the North Sea-Baltic Sea with a horizontal resolution of 1 nautical mile and 50 vertical layers. To account for large scale variations and remotely generated storm surges, the North Sea-Baltic Sea model was nested into a depth-averaged storm surge model of the North Atlantic with a resolution of 5 nautical miles. The atmospheric forcing was derived from the operational model of the German Weather Service with a spatial resolution of 7 km and temporal resolution of 3 hours. A more detailed description of the model system is given by [Gräwe et al. (2015)]. The flow fields for May 2009 were taken out of a running simulation covering the period 1948-2015. The
velocity field was interpolated to an equidistant spacing of 1 m and finally averaged over the upper 10 m to produce a “quasi”
two-dimensional field. The temporal resolution was set to one hour to resolve for instance inertial oscillations.

We have calculated $M_V$ for 11 May 2009 1:00 am with $\tau = 36$ h and applied the eddy tracking based on $M_V$. A $\tau$ value of
36 h corresponds to 15 km and 25 km. One half of the eddies rotates clockwise and the other half counter-clockwise. Their
position in space is drawn from an uniform distribution. To generate a dimensionless model, lengths are measured in units
of the maximal eddy radius of the mesoscale eddies and time in units of the lifetime of an eddy which is chosen equally for
all eddies ($T_e = 1$). $\tau$ is chosen as 0.15. A detailed description can be found in the supplemental material to this article% of
an eddy lifetime of approx. 10-12 days, which was reported previously by [Fennel (2001)]. In contrast to the test case of the
vortex street, we do not expect that the eddies are perfectly circular. To account for deformed and distorted eddies, we had to
introduce a threshold for the convexity deficiency to eliminate contours that are only made out of filaments and are no eddy in
the sense of oceanography. We set the threshold to 11% difference between the area of the convex hull of the points that form
the boundary and the area enclosed by the boundary itself normalised to the area enclosed of the boundary. This definition
of convexity deficiency is according to [Haller et al. (2016)]. Please note, that we still allow to detect contours that cover eddy
merging and decay processes, which are characterized by filaments.

Fig. 11 shows the eddy boundaries detected with the method based on $M_V$ (red) and the ETTB by [Nencioli et al. (2010)] (black)
at 11 May 2009 1:00 am for the same search window size. There are several differences between the number and shapes of
eddies which has to be explained. 150 eddies can be detected with the method based on $M_V$, whereas the ETTB detects only
24 eddies at the same instant of time. One reason for the differences is that $M_V$ contains the information of the velocity field
of a time interval, namely 11 May 2009 1:00 am ±36 h. Each eddy that exists, starts to arise, merges with another eddy or
dies within this time interval leaves a footprint in $M_V$ like the many of small eddies visible in $M_V$. How strong this footprint
is visible in $M_V$ depends on the choice of $\tau$. Therefore, the number of eddies detected with the method based on $M_V$ has to
be compared with the number of eddies detected with the ETTB [Nencioli et al. (2010)] in the whole time interval that is
covered by $M_V$.

The result is presented: The black dots in Fig. 11 As expected the eddy shapes are larger in case of $M$ and $M_V$ than in case
of the eddy tracking tool by [Nencioli et al. (2010)] since both approaches use different definitions of the eddy shape. A problem
arises in case of $M$ and are the eddy cores detected with the ETTB by [Nencioli et al. (2010)] within the time interval 11 May
2009 1:00 am ±36 h. In total, 339 eddies are detected which exist between less than one hour and 72 hours. For some eddies
we will discuss exemplarily why they are detected by one of the methods and not by the other to illustrate which different
problems have to be taken into account if one interprets the results of the different methods.

Close to or within the eddy 1, 2 and 3 detected by the tracking based on $M_V$ if the eddy boundaries are too close to each other
(Fig. 11 upper right eddies). The eddy tracking are several eddy cores detected by the ETTB by [Nencioli et al. (2010)] if one
takes into account the whole time interval. At 11 May 2009 1:00 am the ETTB does not detect eddy 1, 2 and 3 because they
are to weak or do not exist yet. By contrast, the eddy detection method based on $M_V$ detects them due to the construction
of $M_V$ as an integral over time. For eddy 4 only a few eddy cores are detected by the ETTB by [Nencioli et al. (2010)] for the
whole time interval, probably the eddy is to weak and lives to short to be seen as a structure in $M_V$. In case of eddy cores and
identifies the eddies as separated. However, the filaments of the 5 the method based on $M_V$ does not detect an eddy although
the ETTB by Nencioli et al. (2010) detects several eddy cores in the region. One reason could be that the eddy arises, moves
a lot and dies within the time interval such that $M_V$ only captures a blurred structure of the eddy boundary of one eddy are
close to the boundary of the other eddy. The criterion to identify an eddy that does not fulfil the convexity criterion. Eddy 6 is not
detected by the method based on $M_V$, although the eddy boundary is to detect the largest closed contour line with the local
lowest level of $M$ within a search window. If the boundaries are close to each other any lowest local level of $M$ within obvious
in the structure of $M_V$. The reason is that the choice of the search window size for the eddy core detection determines if an
eddy core is detected or not. An enlarged search window could solve this problem for eddy 6, but a larger search window
influences the number of detected eddies. A solution could be an eddy core search independent of the search window can be
clearly identified size.

A general problem, which arises when using surface velocity fields, is that this velocity field is not divergence free. Although,
we have checked that the vertical velocity is small compared to the horizontal ones, there is still a finite residual left. However,
note that this test case is rather artificial. Eddies are placed randomly in the ocean and eddy-eddy interaction is not taken
into account. In realistic applications, if eddies are too close to each other, they certainly will interact, merge, or repel each
other. Hence, the eddy-tracking algorithm might still give reliable results. We still assume that the velocities are two dimensional.

Applying the ETTB by Nencioli et al. (2010) to these “quasi” 2d fields does not cause difficulties, since the algorithm works on
an instantaneous snapshot - a frozen velocity field. Thus, the error made by the 2d assumption is small. The situation changes
when employing a Lagrangian descriptor. During the integration interval $[t^* - \tau, t^* + \tau]$, $M_V$ accumulates these residuals.
Therefore, $M_V$ can show structures that seems to be eddies but are regions of a stronger vertical velocity or Lagrangian
divergence (Jacobs et al. 2016). Therefore, the number of eddies of both methods will include false positives.

In summary, the method based on the combination of the Lagrangian descriptor $M$ and the Euler-Lagrangian-Lagrangian
descriptor $M_V$ can be used for the detection of eddy boundaries that are acting as boundaries of a trapping region. Comparing
the latter to boundaries detected with the method ETTB by Nencioli et al. (2010) leads to large differences in the shape
computed by both methods. These differences and in the size. Those deviations are due to the difference in the definition of
the boundary and yields in case of the vortex street much smaller sizes of the eddies in case of the eddy tracking tool ETTB
by Nencioli et al. (2010). Another advantage of the method based on the combination of the Lagrangian descriptor $M$ and
the Euler-Lagrangian-Lagrangian descriptor $M_V$ is that it even shows filament structures of the eddy boundary in contrast to
the method by Nencioli et al. (2010) ETTB by Nencioli et al. (2010) visible in the example of the western Baltic Sea. These
filaments can be linked to the dynamics of the eddy, e.g. as it starts interacting, merging, or repelling with other eddies or
fading out. Nevertheless, these filamental shapes of eddies might not be eddies according to a more strict mathematical
definition of an eddy boundary as in Branicki et al. (2011); Haller et al. (2016), but they are still important structures in the
flow from an oceanographic point of view and should be considered in a census of eddies.

Nevertheless, one has to take into account that the detection of eddy shapes by the method based on the combination of the
Lagrangian descriptor $M$ and the Euler-Lagrangian-Lagrangian descriptor $M_V$ is restricted by the unlikely case of type 3 noise
in velocity fields.
choice of $\tau$. In highly dynamical velocity fields like the example of the Baltic Sea not all structures can be resolved by the same $\tau$ which leads to a compromise for $\tau$. This choice of $\tau$ influences if an eddy can be detected by the method based on $M_V$ and not by the ETTB by Nencioli et al. (2010) or the other way round.

The method to detect shapes should be chosen based on the question which type of shapes one is interested in and the results of the method should be handled with care.

5 Discussion and conclusion

We have shown, that the introduced heuristic Euler-Lagrangian-Lagrangian descriptor $M_V$ provides a good insight in the flow based on the modulus of vorticity provides good insights into the structure of a hydrodynamic system. It identifies elliptic fixed points that can be linked to eddy cores and, in contrast to Eulerian measures, it can be used to identify eddy cores as well as distinguished hyperbolic trajectories. Eddy cores can be found as local maxima of $M_V$ even identifies hyperbolic fixed points. Moreover, it can distinguish between elliptic and hyperbolic fixed points—a property which cannot be achieved by while DHTs correspond to minima of $M_V$. Hence, compared to the Lagrangian descriptor $M$, therefore, the Euler-Lagrangian descriptor $M_V$ is a suitable tool for tracking eddies and subsequently estimating their lifetimes. Like the Lagrangian descriptor $M$, $M_V$ displays based on the arc length, it does not need an additional criterion to distinguish between eddy cores and DHTs. As any other Lagrangian descriptor it displays singular lines that can be linked to the stable and unstable manifolds and provides as an add on estimates of the eddy boundaries. This fact linked to the idea that manifolds cannot be crossed by trajectories yields a different approach to the tracking of eddy shapes compared to an Eulerian method. Eulerian methods like the one of Nencioli et al. (2010) define the eddy boundary as a line with special properties of the velocity which cannot be linked to the idea of trapping of fluid parcels inside the eddy-DHTs which allows for a simultaneous estimate of the boundaries of the eddies to get an assessment of their size and shape. These features make the quantity $M_V$ suitable for designing an eddy tracking tool, which should be able to detect eddy cores, to track them over time, and additionally to provide information about the eddies’ lifetime, size and shape. Moreover, the eddy tracking should be robust with respect to velocity fields corroborated with errors in case the velocity field is extracted from observations.

We have demonstrated using some paradigmatic velocity fields that the proposed Euler-Lagrangian descriptor $M_V$ yields good results in estimating the To test all those properties in practice we have first used some velocity fields, which are constructed in such a way that the lifetimes of eddies and in combination with are given analytically. It turns out, that the Lagrangian descriptor $M$ eddy-shapes. Therefore, it is a suitable tool for eddy tracking in oceanic velocity fields. The Euler-Lagrangian descriptor $M_V$ gives an estimate of the lifetime that is closest to the analytical lifetime (in the case of the vortex street), therefore it performs much better than any of the other methods in the comparison. It finds eddies earlier than the other methods because is superior in estimating lifetimes compared to the other considered methods. This is due to its definition as an integral which takes the history into account. Eulerian methods like Okubo-Weiss or the ETTB by Nencioli et al. (2010) detect eddies too late and underestimate their lifetime. The formulation of $M_V$ contains information of the future and the past of the velocity field. This property can be useful if one is interested in the process of the eddy birth and its early evolution as an integral is also beneficial
in case of different types of noise. However, there are some disadvantages of this method which need to be addressed. None of the tested methods can deal in a convincing way with type 3 noise which mimics errors to shifts in georeferencing. First of all, it is a heuristic method, that lacks objectivity. This can be problematic since it might lead to failures in the detection of some eddies. But it is easy to implement and fast, so that it compensates for this problem, when this measure is used for computing census and size distributions of eddies in oceanic flows over a long period of time. A general problem of the Lagrangian and the Euler–Lagrangian descriptors is that the resolution of the structures to be detected depends on the chosen time $\tau$. Structures that live too short in relation to the chosen $\tau$ cannot be resolved and will be missed. Hence the choice of $\tau$ contains a decision which time scale and subsequently which eddy lifetime will be resolved. We have shown that the Euler–Lagrangian descriptor

The example of the velocity field of the western Baltic Sea shows that eddy tracking based on $M_V$ provides still good results if the velocity field is corroborated with noise of type 1 and 2. This is due to the smoothing effect inherent in the definition of $M_V$ as an integral along the trajectory. The estimates of the eddy lifetime by $M_V$ is in almost all cases better than the estimate of the other methods. In case of velocity fields with strong type 3 noise collecting the history of the velocity field can be a drawback. As discussed in Sect. 4.2 type 3 noise leads to a divergence of neighbouring trajectories, which disturbs the structure of able to detect the essential eddies that are visible in the velocity field and also detected by the ETTB by Nencioli et al. (2010). Furthermore, it detects eddies that cannot be detected by the ETTB at this instant of time, but was or will be detected by the ETTB at a earlier or later instant of time within the time interval $[t^*-\tau, t^*+\tau]$. Nevertheless, one has to be aware that both ETTB and the eddy tracking based on $M_V$ and in this way the detection of the eddy core or shape. A solution for this problem is the choice of a smaller time-give false positives. The reason could be that structures of strong vertical velocity are identified as eddies. On the other hand false negatives can arise if (i) the eddies are too weak or (ii) the chosen $\tau$ to reduce the time span where the trajectories can diverge from each other. However, as discussed in Sect. 4.2 type 3 noise is very unlikely and completely absent, if numerical generated velocity fields are used value is too large or too small or (iii) the search window is too large or too small.

In general, the choice of the detection method depends on the questions asked. If one is only interested in tracking eddy cores Eulerian methods are a good choice. Lagrangian and especially the Euler–Lagrangian method. By contrast, Lagrangian methods give a more detailed view on the dynamics. Besides, they and provide a more physical estimate of the eddy size. Especially this feature, which describes the fluid volume trapped in an eddy promises to be more useful for applications that consider the growth of plankton populations in oceanic flows. For the latter it has been shown that eddies can act as incubators for plankton blooms due to the confinement of plankton inside the eddy (Oschlies and Garçon (1999); Martin (2003); Sandulescu et al. (2007)).

Author contributions. Rahel Vortmeyer-Kley developed the idea of the eddy tracking tool based on the Lagrangian descriptor $M_V$ and implemented it. Ulf Gräwe supervised the oceanic questions of this work and provided the velocity field for the western Baltic Sea. The overall supervision was done by Ulrike Feudel. All authors contributed in preparing this manuscript.
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Figure 1. Colour-coded representation of the modulus of vorticity (a), Okubo-Weiss-Parameter (b) for the eddy field in Eq. (6). All plots are normalized to the maximum value.


Figure 2. Colour-coded representation of the Lagrangian descriptor $M_V$ (a)-(c) and the Lagrangian descriptor $M$ (d)-(f) for the eddy field in Eq. (6) with $\tau$ chosen as 0.5 ((a) and (d)), 25 ((b) and (e)) and 100 ((c) and (f)). All plots are normalized to the maximum value.
Figure 3. Modulus of vorticity (a), Okubo-Weiss parameter (b), Lagrangian descriptor $M$ (c) and Lagrangian descriptor $M_V$ (d) for the hydrodynamic model of a vortex street at $t = 0.151$ with $\tau = 0.15$, normalized to the maximum value. Blue colours indicate small values yellow large values of the depicted quantity. The dark blue regions in (c) and (d) are regions where the trajectories have left the region of interest.

Figure 4. Eddy lifetime estimated with Okubo-Weiss (OW, violet), modulus of vorticity (absVorticity, cyan), $M_V$ (red) and the eddy tracking tool box (ETTB, blue) by [Nencioli et al., 2010] for vortex strength $w$ 50, 100 and 200. The black diagonal depicts the analytical lifetime.
Figure 5. Time of birth of an eddy estimated with Okubo-Weiss (OW, violet), modulus of vorticity (absVorticity, cyan), $M_V$ (red) and the eddy tracking tool box (ETTB, blue) by [Nencioli et al. (2010)] for vortex strength $w$ 50, 100 and 200. The black diagonal depicts the analytical time of birth of an eddy.

Figure 6. Measured lifetime of an eddy obtained by means of $M_V$ (red line) versus the chosen $\tau$ (analytical lifetime $T_c = 1$ (blue line), vortex strength 200).
Figure 7. Measured median lifetime obtained by different methods (Okubo-Weiss (OW, violet), modulus of vorticity (absVorticity, cyan), $M_V$ (red) and the eddy tracking tool box (ETTB, blue) by Nencioli et al. (2010)) depending on the noise level. The computations have been performed in a velocity field mimicking a vortex street with added white Gaussian noise (type 1 noise with 1000 noise realizations). The errorbars indicate the whiskers of the distribution in the boxplot (not shown here) corresponding to approximately $\pm 2.7\sigma$.

Figure 8. Measured median lifetime obtained by different methods (Okubo-Weiss (OW, violet), modulus of vorticity (absVorticity, cyan), $M_V$ (red) and the eddy tracking tool box (ETTB, blue) by Nencioli et al. (2010)) depending on the noise level. The computations have been performed in a velocity field mimicking a vortex street with type 2 noise (1000 noise realizations). The errorbars indicate the whiskers of the distribution in the boxplot (not shown here) corresponding to approximately $\pm 2.7\sigma$. 
Figure 9. Measured median lifetime obtained by different methods (Okubo-Weiss (OW, violet), modulus of vorticity (absVorticity, cyan), $M_V$ (red) and the eddy tracking tool box (ETTB, blue) by [Nencioli et al., 2010]) depending on the noise level. The computations have been performed in a velocity field mimicking a vortex street with type 3 noise (1000 noise realizations). The errorbars indicate the whiskers of the distribution in the boxplot (not shown here) corresponding to approximately $\pm 2.7\sigma$.

Figure 10. Eddy boundaries detected with the method based on $M_V$ (red line) and with the eddy tracking tool by [Nencioli et al., 2010] (black line) at $t = 0.201$. (a) $M_V$ without noise, (b) $M_V$ with type 1 noise of noise level 0.95, (c) $M_V$ with type 2 noise of noise level 0.95, (d) $M_V$ with type 3 noise of noise level 0.5. The $\tau$ value is chosen as 0.15 $T_c$ with $T_c = 1$. The dark blue regions are regions where the trajectories have left the region of interest. All plots are normalized to the maximum value.
Figure 11. $M_V$ for the western Baltic Sea for 11 May 2009 1:00 am with $\tau = 36$ h normalized to the maximum value of $M_V$. The red lines are the eddy boundaries and red dots the eddy cores detected with the method based on $M_V$ respectively. The black lines are the eddy boundaries detected with the ETTB by Nencioli et al. [2010] at 11 May 2009 1:00 am. The black dots are the eddy cores detected with the ETTB by Nencioli et al. [2010] within the time interval 11 May 2009 1:00 am $\pm$ 36 h. The dark blue regions are areas where the trajectories have left the domain of interest, light grey regions indicate land.
Supplemental material of: Detecting and tracking eddies in oceanic flow fields: A vorticity-Lagrangian descriptor based Euler-Lagrangian method on the modulus of vorticity

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S1 Seeded eddy model

The model is inspired by the seeded eddy model of a turbulent flow described in [Abraham 1998] and the model of [Jung et al. 1993]. The model consists of n eddies and a background flow u₀ in positive x-direction. The stream function for an eddy field with n eddies is given by:

\[ \psi(x, t) = \sum_{i=1}^{n} (-1)^{m_i} \cdot w_i \cdot \exp(-\kappa_i ((x - x_i)^2 + \alpha (y - y_i)^2)) + u_0 \cdot x \]

where \( \kappa_{i}^{-1/2} \) is linked to the radius of eddy \( i \) with vortex strength \( w_i \). The eddy \( i \) emerges at position \( (x_i, y_i) \). The radii of the eddies are drawn from a distribution of 1000000 eddies possessing a power law \( 1/r^3 \). The distribution is normalized and allows eddies between \( R_{min} = 10 \text{ km} \) and \( R_{max} = 50 \text{ km} \). Because we are interested in mesoscale eddies (15-25 km radii) we first draw a sample of 100 radii from the distribution and then pick randomly 10 sample radii that are between 15 km and 25 km. The position of these eddies is drawn from an uniform distribution. Half of the eddies rotates counter-clockwise and the other half clockwise. The factor \( m_i \) decides if the eddy rotates clockwise or counter-clockwise. If \( m_i \) is even the eddy rotates clockwise if \( m_i \) is odd the eddy rotates counter-clockwise. The model is parametrized such that it yields a dimensionless model. Lengths are measured in units of the maximal eddy radius of the mesoscale eddies and time is in units of the lifetime of an eddy which is the same for all eddies. The parametrization is given in Table ??.

<table>
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<th>Dimensionless</th>
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</tr>
<tr>
<td>( T_c )</td>
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</tr>
<tr>
<td>( \alpha )</td>
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</tr>
<tr>
<td>( \kappa )</td>
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<tr>
<td>( u_0 )</td>
<td>55 m²/s⁻¹</td>
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11 11 18.66 200
**Table S1.** Dimensional and dimensionless parameters of the hydrodynamical model of a vortex street

<table>
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<th>Parameter</th>
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<th>$T_c$</th>
<th>$\alpha$</th>
<th>$\kappa$</th>
<th>$L$</th>
<th>$u_0$</th>
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<td>1</td>
<td>1</td>
<td>1</td>
<td>6</td>
<td>18.66</td>
<td>0.5</td>
<td>200</td>
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S2  The hydrodynamical model of a vortex street

The hydrodynamical model of the vortex street is an adapted version of the model by Jung et al. (1993). The model is modified in such a way that two counter rotating eddies that develop at times $t$ and $t + T_c/2$ at position $(1, y_0)$ and $(1, -y_0)$ respectively, travel a distance $L$ in positive $x$-direction within their lifetime $T_c$ and fade out. The model is artificial because the impact of the cylinder on the eddy formation and its shading are neglected here making the eddy formation non-physical out of the blue nowhere. However, since all quantities to be estimated by the eddy tracking tool are then given analytically, makes this artificial model makes up an ideal test bed for numerics.

Hence, the model is simplified as follows:

$$
\Psi(x, y, t) = -wh_1(t)g_1(x, y, t) + wh_2(t)g_2(x, y, t) + u_0 y. \tag{1}
$$

The first two terms describe the life cycle of the eddies of opposite sense of rotation. The vortex strength is given by $w$. The dynamics of the eddy evolution is modelled as the modulation of the amplitudes by the function $h_1(t) = |\sin(\pi t/T_c)|$ resp. $h_2(t) = h_1(t - T_c/2)$. The function $g_i(x, y, t) = \exp(-\kappa_0((x - x_i(t))^2 + \alpha(y - y_i(t))^2))$ with $i = 1, 2$ models the Gaussian shaped effect of the eddies on the stream function. The movement of the eddy centres is expressed by $x_1(t) = 1 + L(t/T_c \mod 1), x_2(t) = x_1(t - T_c/2)$ and $y_1(t) = y_0 = -y_2(t)$. (In case of type 3 noise, the noise is added to the $y_1$ and $y_2$ terms.) The factor $\kappa_0^{-1/2}$ is the radius and determined as a characteristic linear size of the eddies and $\alpha$ is the ratio between the elongations of the eddy in $x$ and $y$ direction. In our case $\alpha$ is set to 1 (circular eddies). The last term describes the background flow with the velocity $u_0$.

The parametrization of the flow is chosen as in Sandulescu et al. (2006). Lengths are measured in units of the eddy radius $r$ and time in units of the lifetime $T_c$ of an eddy. The dimensional and dimensionless parameters are given in Table [S1] The vortex strength is furthermore varied to study its impact on the eddy evolution.

S2  Algorithm of the eddy tracking

The key ideas of the algorithm (eddy detection, eddy tracking and eddy shape) inspired by the eddy tracking package by Nencioli et al. (2010) are schematically presented in this section.
Fig. [S1] presents the concept of eddy detection. Similar algorithms can be constructed for the modulus of the vorticity and for the Okubo-Weiss criterion (with respect to taking into account the fact that eddy cores are minima in case of Okubo-Weiss).

Fig. [S2] shows idea of the eddy tracking.

Fig. [S3] presents the eddy shape detection based on the combination of the Euler-Lagrangian modulus of vorticity based Lagrangian descriptor $M_V$ and the Lagrangian descriptor $M$ based on the modulus of vorticity. The eddy shape detection searches for the largest closed contourline with the largest gradient of $M_V$ along the contour line. To maximise these two conditions at the same time, we maximise a quantity that combines this two ideas: \( \text{(Area enclosed by the contour line)} \times \left( \sum \text{gradient of } M_V \text{ along contour line} \right) / \left( \text{length of contour line} \right) \). Because contour lines surround the eddy core like the layers of an onion, maximising the enclosed area includes maximising the length of the closed contourline. Maximising the gradient of $M_V$ along the contourline is linked to the idea to search for a singular line (the eddy boundary).

In case of eddy shape detection for realistic oceanic velocity fields like the example of the western Baltic Sea, the coordinates of the eddy cores have to be understood as candidates for the eddy core.

From all candidates for eddy cores only those are kept which fulfil the following conditions:

- The convexity deficiency as defined in [Haller et al. (2016)] has to be smaller than a threshold.
- No land is enclosed in the contourline.
- The contourlines are longer than a threshold length. The reason for that criterion is mainly to speed up the computation. Very short contourlines can typically be found very close around the eddy cores. Therefore, they need not to be checked for the gradient of $M_V$, because they will not describe the eddy boundary.

This set of eddies can then be used as input for eddy tracking.
**Figure S1.** Schematic sketch of the eddy detection algorithm based on Euler-Lagrangian descriptor $M_V$.

**References**


Figure S2. Schematic sketch of the eddy tracking algorithm based on **Euler-Lagrangian Lagrangian** descriptor $M_V$.

Figure S3. Schematic sketch of the eddy shape algorithm based on the combination of the Euler-Lagrangian descriptor $M_V$ and the Lagrangian descriptor $\delta M_V$. 