

Eddy diffusivity in the Gulf of Finland: Dispersion model evaluation based on drifter data

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Particle tracking by Lagrangian Stochastic models

Lagrangian Stochastic models are used to track the movement of individual particles, which may represent a fluid element, a solute or a distinct, insoluble substance.

- 0th-order LSM:

$$dX_i = (U_i + \partial K_i / \partial x_i) dt + \sqrt{2K_i} d\xi_i$$

X_i = particle position; U_i = mean Eulerian velocity

K_i = eddy diffusivity; $d\xi_i$ = random forcing

- *Wiener process*: $d\xi_i$ determined by a Gaussian distribution with zero mean and variance equal to time increment dt .
- The 0th-order LSM assumes that the time increment dt is large compared to the integral time scale of turbulence.

Q1 : What are appropriate values for K_i ?

Particle tracking by Lagrangian Stochastic models

- 1st-order LSM:

$$dX_i = (U_i + u_i)dt$$
$$du_i = a_i(\vec{X}, \vec{U}, t)dt + b_{ij}(\vec{X}, \vec{U}, t)d\xi_j$$

where u_i = velocity fluctuations

- If the autocorrelation function decays exponentially: $R(t) = \exp(-t/T_v)$, then a_i and b_{ij} are determined as

$$a_i = -\frac{u_i}{T_v} \quad b_{ij} = \sqrt{\frac{2\sigma_u^2}{T_v}}$$

- The 1st-order LSM retains memory of its initial velocity for a time T_v .

Q2 : Is the 0th-order LSM sufficient for simulations in the Gulf of Finland, or is a 1st-order LSM required?

Gulf of Finland

- Estuarine shallow sea (mean depth: 38 m)
- Surface currents dominated by wind forcing and river runoff.
- High risk area for pollution due to ship accidents
 - Intense ship traffic both along the main east-west axis of the GoF, crossing with traffic between Tallinn and Helsinki

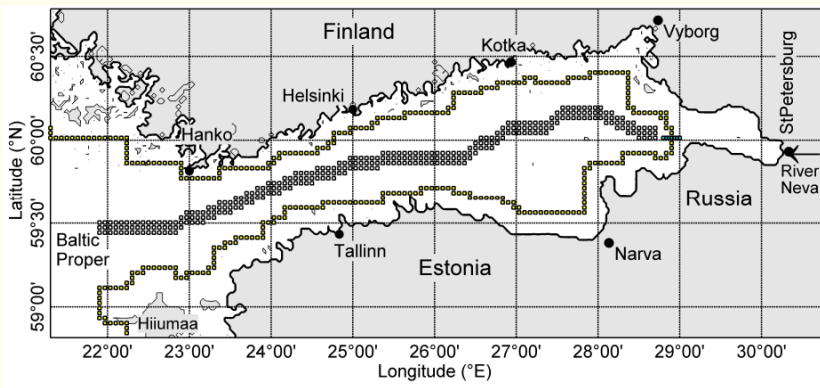


BalticWay: Potential for coastal pollution from shipping

(Viikmäe et al. 2013, Ocean Dynamics, **63**(5), pp 589–597)

Potential for coastal pollution from shipping analyzed based on simulations (TRACMASS) of Lagrangian particle trajectories starting from the fairway.

- How does particle diffusion influence the simulation results?



Map of the Gulf of Finland with near-shore points and major fairway.

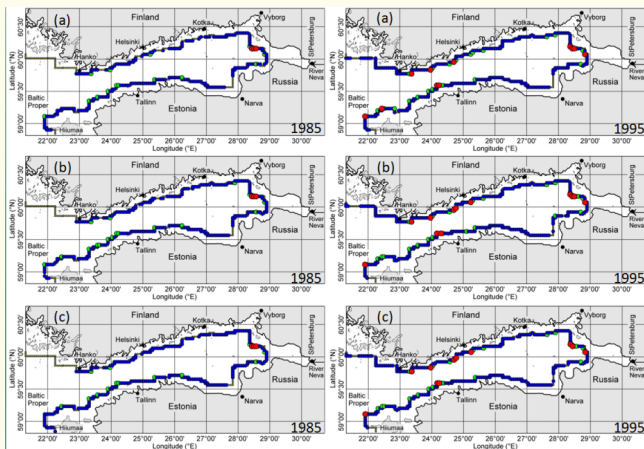
BalticWay: Potential for coastal pollution from shipping

Annual mean distribution of particle hits along the coast.

Left panels: 1985

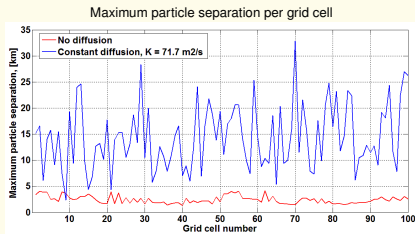
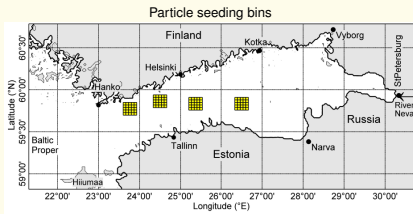
Right panels: 1995

- (a) No particle diffusion
- (b) Constant particle diffusion coefficient
- (c) Particle diffusion with Smagorinsky formula



- Particle diffusion adds a small number of hits (about 3%) when compared with simulations without particle diffusion.

Effect of particle dispersion



- Viikmäe et al. 2013: Constant eddy diffusivity coefficient $K = 5 \text{ m}^2/\text{s}$, suggested by Döös and Engqvist for Baltic Sea (Estuar. Coast. Shelf. Sci. 74:585–597, 2007).
- De Dominicis et al. (Ocean Dynamics **62**(9), pp 1381–1398) found $K \sim 1 \times 10^3 \text{ m}^2/\text{s}$ for surface drifters in the Adriatic Sea and Ligurian Sea.

The appropriate value of K for Gulf of Finland should be determined by local surface drifter experiments.

Surface drifters

Design: GPS/GPRS tracker + battery pack + 2m long plastic tube

Battery life: approximately 4 weeks



GPS tracker,
communication
by GPRS/GSM
network



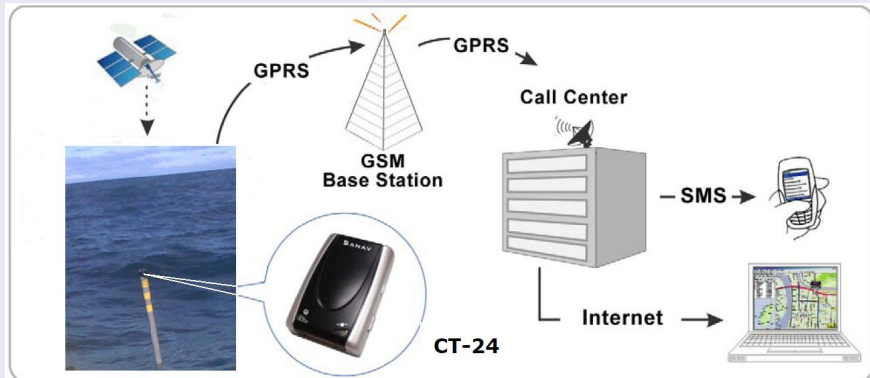
Sanav CT-24
(replaced by
MU-201 in new
design)



2013: Updated slender design above
sea level, minimize wind drag

Surface drifters

Drifter data flow



Complexity of single deployment tracks

- Single drifter trajectories display chaotic behaviour.
- Pairs/clusters of drifters deployed close to each other ($r \sim 50\text{m}$) typically stay together for several days.
- Drifters report position every 10 or 15 min.
- Connection errors may lead to gaps in drifter records. Short gaps (~ 3 hours) in drifter positions can be filled by linear data interpolation.

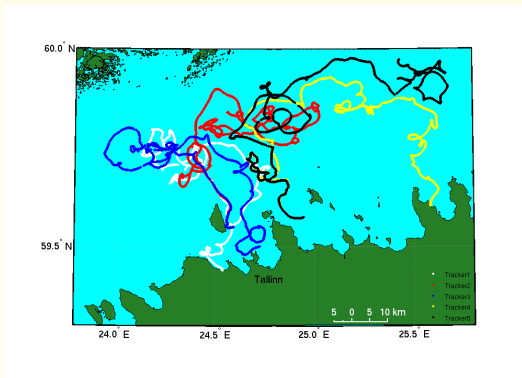


Figure: Deployment of 5 drifters, 10/4/2014

Surface drifter experiments: 2010 – 2014

YEAR	single	pair	triplet	quadruplet	total
2010	2	1	3	-	13
2011	-	1	8	-	26
2012	-	-	-	-	-
2013	-	3	5	1	25
2014	1	3	1	-	10

- Track duration: 1–35 days
- Drifter deployments within a few nautical miles from the Estonian coast.
 - Several short tracks due to drifters brought directly to land.
- Drifters deployed in the period April to October.
 - Most drifters deployed in summer and early autumn.
 - Season dominated by mild wind and wave climate.

Surface drifter experiments: 2010 – 2013

Study area: Western part of Gulf of Finland

Area of extensive ship traffic in crossing shipping lanes, where a potential oil spill is likely to reach either the Estonian or Finnish coast.

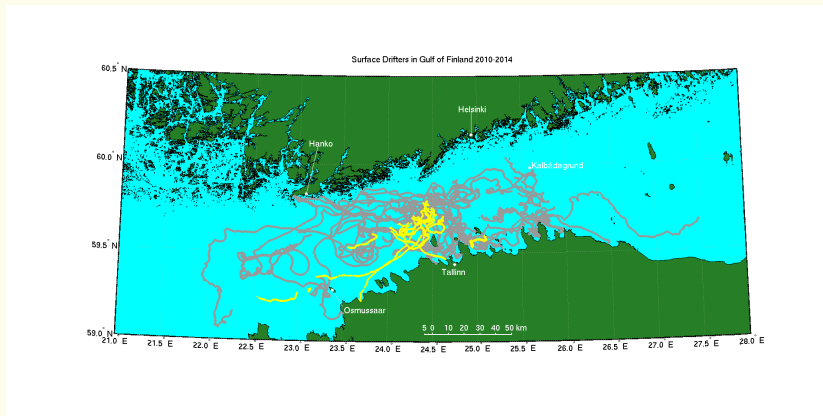


Figure: Surface drifter trajectories: Deployments in 2010

Surface drifter experiments: 2010 – 2013

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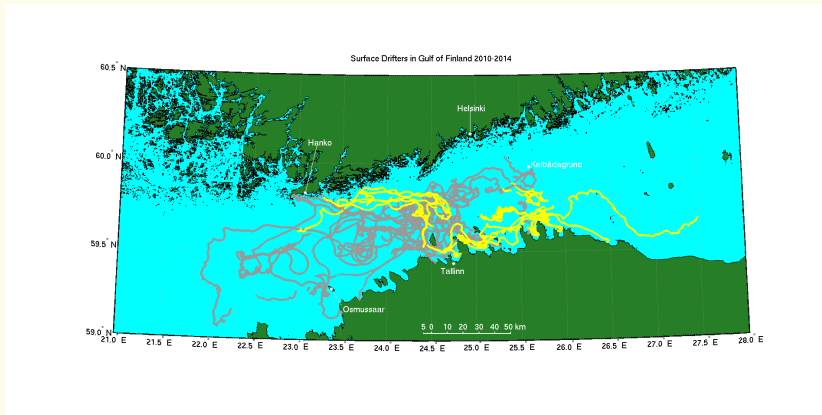


Figure: Surface drifter trajectories: Deployments in 2011

Surface drifter experiments: 2010 – 2013

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Area of extensive ship traffic in crossing shipping lanes, where a potential oil spill is likely to reach either the Estonian or Finnish coast.

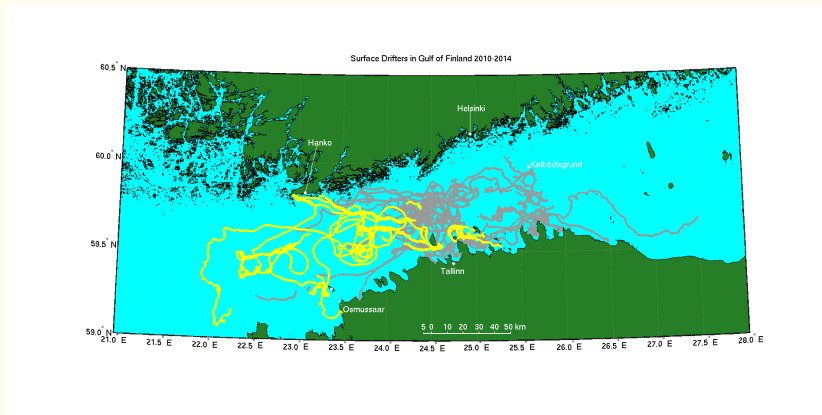


Figure: Surface drifter trajectories: Deployments in 2013

Surface drifter experiments: 2010 – 2013

Study area: Western part of Gulf of Finland

Area of extensive ship traffic in crossing shipping lanes, where a potential oil spill is likely to reach either the Estonian or Finnish coast.

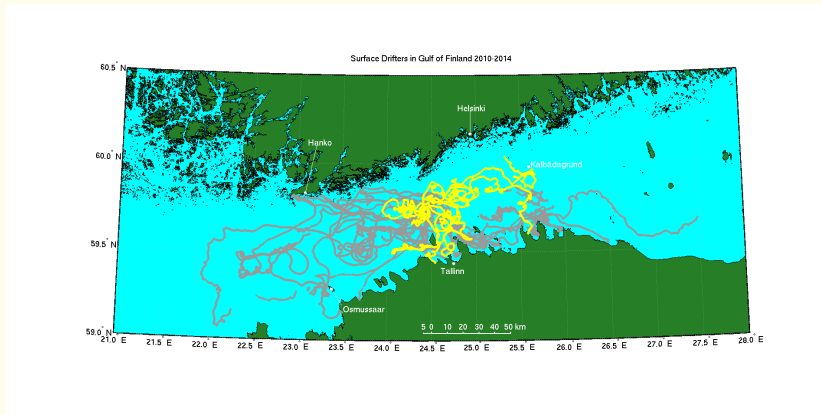


Figure: Surface drifter trajectories: Deployments in 2014

Drifter velocity distribution

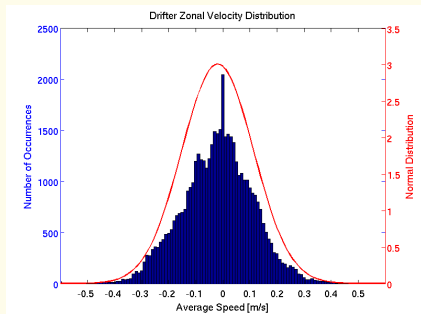


Figure: Zonal velocity distribution

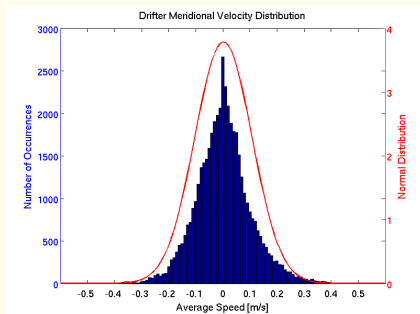


Figure: Meridional velocity distribution

Velocity	Mean [m/s]	SD [m/s]
Zonal	-0.019	0.133
Meridional	0.002	0.105

Table: Mean and Standard Deviation of normal distribution fit for the Zonal and Meridional velocity components.

Drifter velocity distribution

Sample: 46 080 records,
10 min sampling rate

U_{MEAN}	0.143 m/s
$U_{RAYLEIGH}$	0.121 m/s
σ_U	0.092 m/s

Table: Comparison of drifter mean speed and maximum likelihood of Rayleigh distribution fit.

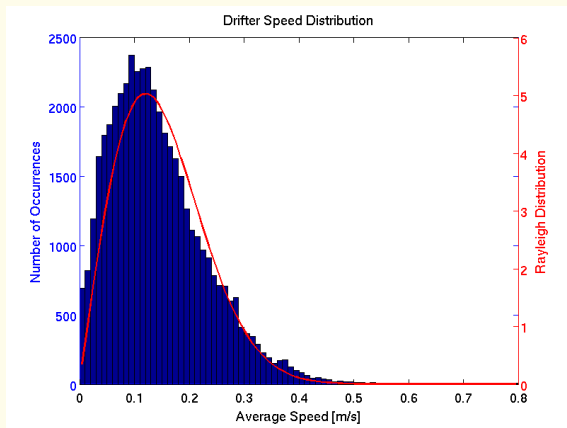


Figure: Drifter speed distribution

Lagrangian Integral Time

A key parameter for single particle analysis is the Lagrangian “integral time” T_L , which depend on the velocity auto-correlation for single trajectories, and provides a basic indicator of Lagrangian predictability.

Calculation of T_L :

- Decomposed velocity components as

$$U_i(t) = \overline{U}_i + u_i(t), \quad i = \{1, 2\} \quad (1)$$

where \overline{U}_i is the time average of $U_i(t)$ and u_i is the time varying deviation from the mean.

- Calculate the auto-covariance of u_i , defined by

$$\sigma_i^2(\tau) = \lim_{T_m \rightarrow \infty} \frac{1}{T_m} \int_0^{T_m} u_i(t + \tau) \cdot u_i(t) dt \quad (2)$$

Lagrangian Integral Time

Calculation of T_L (cont.):

- The auto-correlation function

$$R_i(\tau) = \frac{\sigma_i^2(\tau)}{\sigma_i^2(0)} \quad (3)$$

is the auto-covariance normalized by $\sigma_i^2(\tau = 0)$.

- The zonal and meridional Lagrangian integral time is calculated as

$$T_{L,i} = \int_0^{\infty} R_i(\tau) d\tau \quad (4)$$

- The total Lagrangian integral time T_L is calculated as the mean value of the zonal and meridional components

$$T_L = \frac{1}{2} [T_{L,1} + T_{L,2}] \quad (5)$$

Problems with calculation of T_L for surface drifter data:

1. Finite drifter track duration

- Drifter trajectories have finite duration, which limits the auto-covariance calculation eq. (2) to use $T_m = \text{drifter track duration}$.

$$\sigma_i^2(\tau) = \frac{1}{T_m} \int_0^{T_m} u_i(t + \tau) \cdot u_i(t) dt \quad (6)$$

- Splitting long trajectories into several short trajectories increases the pool of available tracks in the analysis. However, a small value of T_m changes the auto-covariance function σ_i^2 .

Q1: How small can T_m be without changing the value of T_L ?

Lagrangian Integral Time : Problems

Problems with calculation of T_L for surface drifter data:

2. Noise dominates the auto-correlation function $R(\tau)$

The auto-correlation function $R(\tau)$ becomes noisy for $\tau \gg 0$, hence it is customary to truncate the Lagrangian integral time at the de-correlation time scale. Sometimes alternative definitions of T_L are used:

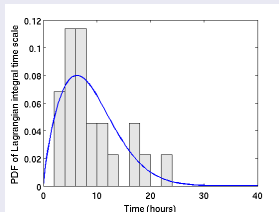
standard formula	“squared auto-correlation”
$T_{L,i} = \int_0^{T_D} R_i(\tau) d\tau$	$T_{L,i} = \int_0^{T_D} R_i^2(\tau) d\tau$

Methods used for calculation of T_L :

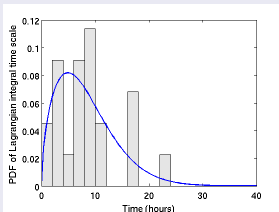
- (A) T_D defined as the first time for which $R_i(\tau) = 0$.
- (B) T_D defined as a constant value: $T_D = 48$ hours.
- (C) Use the “squared auto-correlation” formula with $T_D = 48$ hours.

Q2: What is the most reliable method for calculation of T_L ?

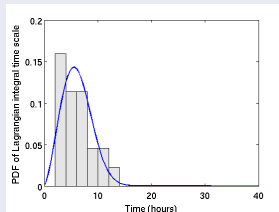
Lagrangian Integral Time : Example



Method A



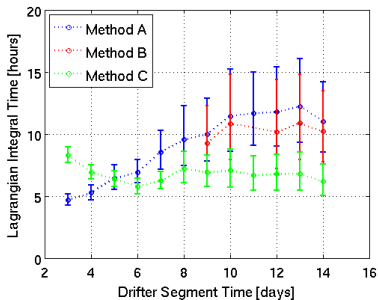
Method B



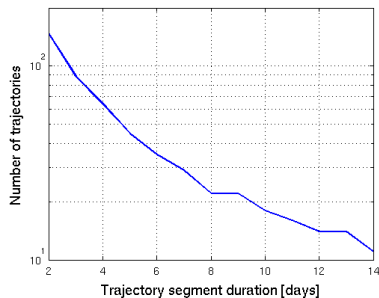
Method C

Lagrangian Integral Time T_L calculated for drifter trajectory segments with duration of 9 days. The blue curve represent data fitting using the Weibull distribution function.

Lagrangian Integral Time : Results



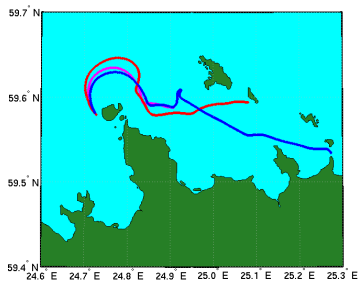
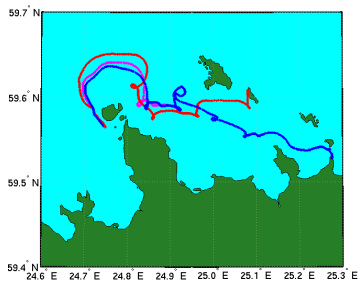
Maximum likelihood estimate for the Lagrangian Integral Time based on a Weibull data fit.



Number of drifter tracks available for analysis depending on track duration T_m .

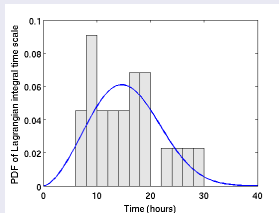
Filtering of inertial oscillations

Surface drifters are subject to inertial and sub-inertial motions, which in the Gulf of Finland have periods $T_{osc} < 14$ hours. These motions are usually filtered out when analyzing the persistency of drifter trajectories, which can be accomplished by applying a running mean of $T = 14$ hours to the sequence of trajectory positions.

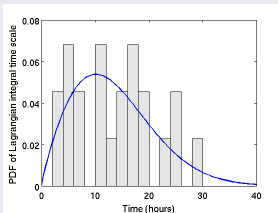


Comparison of original and filtered surface drifter tracks, for surface drifters deployed on 2nd July 2013.

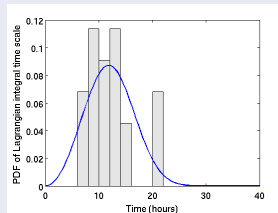
Filtering of inertial oscillations : Example



Method A



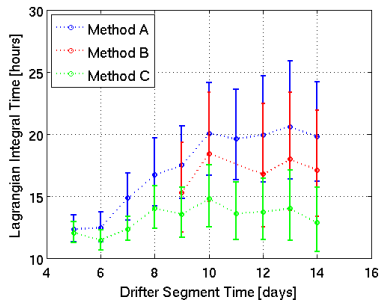
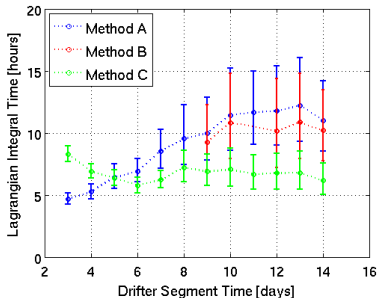
Method B



Method C

Lagrangian Integral Time calculated for drifter trajectory segments with duration of 9 days, **filtered with a 14-hour running mean**. The blue curve represent data fitting using the Weibull distribution function.

Filtering of inertial oscillations : Results



Comparison of T_L for original and filtered surface drifter data: Maximum likelihood estimate for the Lagrangian Integral Time based on a Weibull data fit.

Lagrangian Time and Length scales

Given the Lagrangian integral time scale T_L and the characteristic eddy speed defined by the standard deviation u'_i of the velocity components, we can define the Lagrangian eddy length scale

$$L_L = u' T_L \quad (7)$$

Lagrangian eddy diffusivity: $K = u'^2 T_L$

Method	un-filtered			filtered		
	T_L [h]	L_L [km]	K [m ² /s]	T_L [h]	L_L [km]	K [m ² /s]
A	12 ± 3	3.9 – 6.5	620	20 ± 4	6.9 – 10.4	580
B	11 ± 3	3.5 – 6.0	570	18 ± 4	6.0 – 9.5	520
C	7 ± 1	2.6 – 3.5	360	14 ± 2	5.2 – 6.9	410

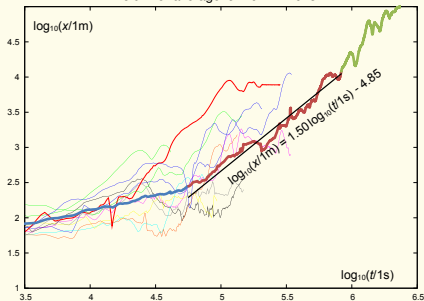
Table: Lagrangian integral time and Lagrangian eddy length scale calculated by methods A, B and C, with $u' \approx 0.12$ m/s ($u' \approx 0.09$ m/s filtered).

Drifter pair analysis: Relative dispersion

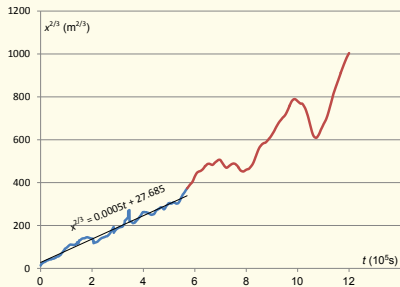
Relative dispersion

Analysis of relative dispersion of drifters is based on the displacement $\mathbf{r}(t)$ between drifter pairs. The relative distance $x(t) = |\mathbf{r}(t)|$ is determined as a function of time elapsed since the moment of release.

(a) Thin lines : Relative distance for deployments in 2011
Bold line: average for 2011 – 2013



(b) Drifter data 2011 – 2013, plotted in power-law scale $x^{2/3}$.



Relative distance x between drifters pairs. Bold curve depicts geometrical average of the pair distance using the data for all the 29 drifter pairs recorded during the years of 2011 and 2013.

Richardson's 4/3 power law

- Relative diffusivity for particle pairs: $K_R = \frac{1}{2} \frac{d}{dt} \langle x^2 \rangle$.
- Richardson (1926) proposed that $\langle x^2 \rangle \propto t^3$
⇒ Scale dependent relative diffusivity: $K_R \propto x^{4/3}$.

Relative drifter velocity $\mathbf{v}(\mathbf{r}, t) = (v_1, v_2)$ is determined by

$$\mathbf{v}(\mathbf{r}, t) = \frac{\mathbf{r}(t_{i+1}) - \mathbf{r}(t_i)}{t_{i+1} - t_i}, \quad t = \frac{1}{2}(t_{i+1} + t_i) \quad (8)$$

- Energy in turbulence inertial range: $E(k) \propto k^{-5/3}$ (Kolmogorov, 1941).
- Richardson's law assumes that the velocity spectrum follows the Kolmogorov law: the fluctuations associated with a distance x should scale as $v = |\mathbf{v}(\mathbf{r}, t)| \propto x^{1/3}$.

Relative dispersion : Velocity spectrum

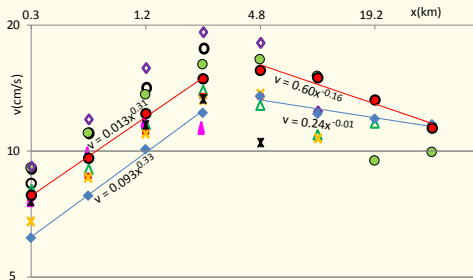


Figure: Drifter velocity fluctuations $v(x)$ are plotted against the associated scale x in log-log plot; Kolmogorov law corresponds to a linear curve with the slope of $1/3$. Different data points correspond to different experiments (with different winds and flow patterns). Linear regression lines are given for two data sets, separately for scales smaller than 4 km and for scales larger than 4 km.

The velocity spectrum calculation based only on data along trajectories becomes non-reliable for large scales due to the twisting of trajectories.

Spatial variability: Drifter velocity

Spatial variability: Drifter velocity

Average drifter velocity [m/s] for bin sizes of about $11 \times 11 \text{ km}^2$. Calculation using all drifter data in period 2010–2014.

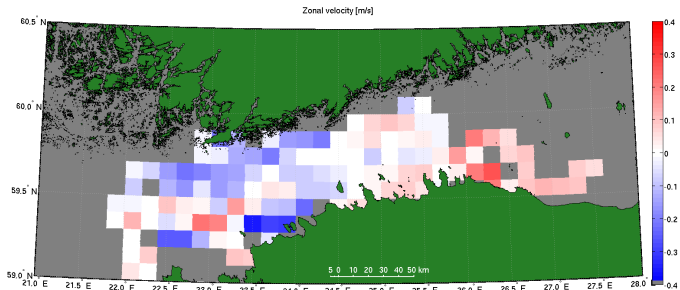


Figure: Zonal drifter velocity

Spatial variability: Drifter velocity

Average drifter velocity [m/s] for bin sizes of about $11 \times 11 \text{ km}^2$. Calculation using all drifter data in period 2010–2014.

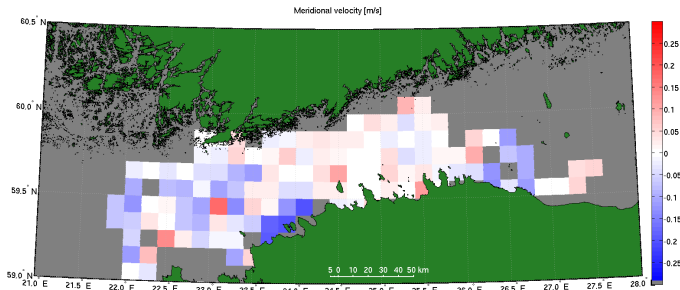


Figure: Meridional drifter velocity

Spatial variability: Drifter velocity

Average drifter velocity [m/s] for bin sizes of about $11 \times 11 \text{ km}^2$. Calculation using all drifter data in period 2010–2014.

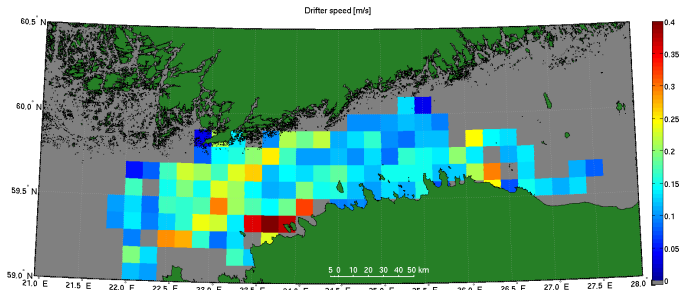


Figure: Drifter speed

Summary & Concluding remarks

- Surface drifters have an average speed $U_{MEAN} = 0.14$ m/s, with extremes up to $U \approx 0.5$ m/s, when calculated from data with 10 min sampling rate.
- Record durations of $T_m \sim 8\text{--}10$ days are required to obtain reliable results for calculation of the Lagrangian integral time T_L .
- The Lagrangian integral time is $T_L \in [7, 12]$ hours, extended to $T_L \in [14, 20]$ hours with filtering of inertial oscillations, depending on which method is used for the calculation.
- At large times $t \gg T_L$, the constant dispersion coefficient is in the range $K \in [360, 620]$ m²/s.
- Relative drifter separation $x(t)$ follows a Richardson's law dispersion characterized by the time dependent relative distance increasing as $x(t) \propto t^{3/2}$. This behavior is a prominent feature for most surface drifter pairs separated by a distance from ~ 50 m up to ~ 5 km, and may extend even beyond this scale.

Summary & Concluding remarks

- Further analysis is required to determine spatial and seasonal variability for surface drifters. For such analysis the problem of insufficient data coverage is a significant factor.
- For a typical LSM simulation in the Gulf of Finland: $dt \ll T_L$. This suggests that a 1st-order LSM should be used for such simulations.

Thank you for your attention!