

Correlation between long-term marine temperature time series from the eastern and western English Channel: Scaling analysis using empirical mode decomposition

Jonathan Derot, François G Schmitt, Valérie Gentilhomme, Pascal Morin

► To cite this version:

Jonathan Derot, François G Schmitt, Valérie Gentilhomme, Pascal Morin. Correlation between long-term marine temperature time series from the eastern and western English Channel: Scaling analysis using empirical mode decomposition. Comptes Rendus Géoscience, 2016, 348 (5), pp.343-349. 10.1016/j.crte.2015.12.001. hal-01285576

HAL Id: hal-01285576 https://hal.sorbonne-universite.fr/hal-01285576

Submitted on 9 Mar 2016

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



Distributed under a Creative Commons Attribution - NonCommercial - NoDerivatives 4.0 International License

C. R. Geoscience xxx (2016) xxx-xxx



Contents lists available at ScienceDirect

Comptes Rendus Geoscience



www.sciencedirect.com

External Geophysics, Climate

Correlation between long-term marine temperature time series from the eastern and western English Channel: Scaling analysis using empirical mode decomposition

Jonathan Derot ^{a,b}, François Guillaume Schmitt ^{a,*}, Valérie Gentilhomme ^c, Pascal Morin ^{d,e}

^a UMR LOG 8187, laboratoire d'océanologie et de géosciences, université du Littoral Côte d'Opale, 32, avenue Foch, 62930 Wimereux, France

^b UMR LOG 8187, CNRS, laboratoire d'océanologie et de géosciences, 28, avenue Foch, 62930 Wimereux, France

^c UMR LOG 8187, laboratoire d'océanologie et de géosciences, université de Lille-1, 28, avenue Foch, 62930 Wimereux, France

^d UMR AD2M 7144, équipe chimie marine, station biologique de Roscoff, place Georges-Teissier, 29680 Roscoff, France

^e Sorbonne universités, UPMC, université Paris 06, UMR7144, adaptation et diversité en milieu marin, station biologique de Roscoff, 29680 Roscoff, France

ARTICLE INFO

Article history:

Received 10 December 2015 Accepted after revision 11 December 2015 Available online xxx

Handled by Didier Roche

Keywords:

Temperature dynamics Autonomous monitoring English Channel Cross-correlation analysis Time dependent intrinsic correlation method Power spectra Empirical mode decomposition

1. Introduction

In recent years, there has been an increasing global concern in protecting the biodiversity and water quality of marine and coastal waters. Such a concern is present for example in two key European parliament directives: the Water Framework Directive (WFD, Directive 2000/60/EC of 23 October 2000) and the Marine Strategy Framework Directive (MFSD, Directive 2008/56/EC of 17 June 2008). These have led to a greater number of marine protected

* Corresponding author. *E-mail address:* fg.schmitt@me.com (F.G. Schmitt).

ABSTRACT

High-frequency temperature fluctuations recorded in the English Channel are compared using two long-term autonomous underwater monitoring stations at less than 20-min time resolution. Measurements were taken from 2005 to 2011 from two systems 460 km apart in the western and eastern parts of the English Channel. Spectral analysis reveals similar statistical behaviors, with approximate 5/3 spectra and several forcing frequencies in relation to tidal and daily cycles. A co-spectra study reveals a transition scale of 11 days. The influence of this scale is also visible though Time-Dependent Intrinsic Correlation analysis (TDIC)–a recently introduced cross-correlation analysis based on Empirical mode decomposition. This helps to spatialize high-frequency temporal records at a fixed location.

© 2016 The Authors. Published by Elsevier Masson SAS on behalf of Académie des sciences. This is an open access article under the CC BY-NC-ND license (http://creativecommons.org/licenses/by-nc-nd/4.0/).

areas, including seas, oceans, and lakes, all with the aim of protecting natural or cultural resources.

In the framework of such European directives, monitoring systems in coastal waters is essential in collecting information on physical, biogeochemical, and biological parameters. Many monitoring programs have been installed with weekly or monthly on-site sampling. Due in part to physical forcing, such as waves and turbulence, the marine environment is highly variable on smaller time scales. Consequently, several autonomous and automatic monitoring programs have been installed in recent years (Chang and Dickey, 2001; Chavez et al., 1997; Dickey, 1991; Dickey et al., 1993; Dur et al., 2007; Nam et al., 2005). These operate typically at 10 to 30 min time

http://dx.doi.org/10.1016/j.crte.2015.12.001

1631-0713/© 2016 The Authors. Published by Elsevier Masson SAS on behalf of Académie des sciences. This is an open access article under the CC BY-NC-ND license (http://creativecommons.org/licenses/by-nc-nd/4.0/).

2

ARTICLE IN PRESS

J. Derot et al./C. R. Geoscience xxx (2016) xxx-xxx

resolution at fixed position, corresponding to Eulerian sampling. There is a factor of about 1000 between sampling at 20-min resolution and sampling every two weeks. In the following, 20-min sampling resolution is referred to as "high frequency". There are sensors able to measure physical or biogeochemical quantities at even higher frequencies (Seuront et al., 1999), but such systems have not been operating autonomously for several years.

Here we focus on the temporal variability of temperature using two data sets recorded at fixed locations along the French coastal area of the English Channel. The first one is a mooring, located in the western English Channel in the bay of Morlaix. The second is an automatic system called MAREL Carnot, located in the eastern English Channel at the exit of the harbor of Boulogne-sur-Mer. In both cases, the data from these automatic systems were recorded long-term (such denomination is usually used for record periods larger than about five years) using an automatic system at high frequency, with more than one measurement per hour.

High-frequency measurements at a fixed location correspond to an Eulerian sampling of a highly variable time-scale system. Sometimes such measurements are considered limited due to the fact that they provide a local view in time, where spatial information is lacking. Here, in order to have a view on the temporal and spatial link between two high-frequency measurement time series (recorded at two locations separated by several hundred kilometers), we shall study the correlation between the two series. The study of the multiscale correlations between these time series will provide us with the first hint at a spatialization of a time series at fixed locations. In this context, we will consider seven years of data, between 2005 and 2011, for two different temperature records. The following section presents the characteristics of these automatic systems, their datasets, and their climatology. We then proceed with showing some spectral and cospectral analyses. The last section considers time dependent intrinsic correlation analysis of both time series–a new correlation method based on the empirical mode decomposition.

2. Data presentation and climatology

2.1. Estacade at Roscoff

The first of the two stations is considered here, where measurements of temperature were taken at the coastal site of Estacade (48°43.56 N, 3°58.58 W) located in the bay of Morlaix in the western English Channel (Fig. 1), from 2005 to 2011. In this area, characterized by strong tidal currents and tidal amplitude over 6 m, the water column remains completely mixed throughout the year. Temperature was measured with a sampling frequency of 10 min using a Seabird SBE39 recorder installed on a mooring at the sea bottom (mean water depth: 5.8 m). The percentage of available data for this period is close to 97%, which corresponds to 329,294 data points acquired (Fig. 2). The analytical precision of temperature measurements from the Seabird SBE39 sensor is 0.01 °C; the raw data at the output of this system are presented in Fig. 2 with a green pattern.

2.2. MAREL Carnot at Boulogne-sur-Mer

The second temperature datasets were recorded by an automatic system called MAREL Carnot, which is situated in the eastern English Channel in the coastal waters of

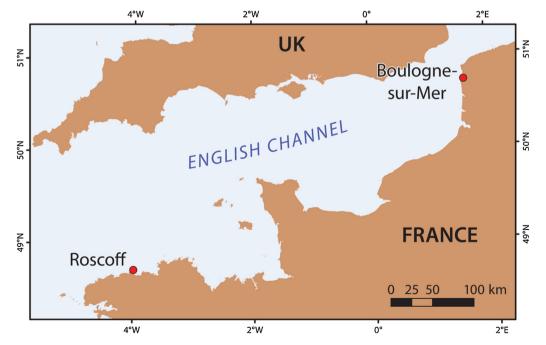


Fig. 1. A map showing the measurement location in Boulogne-sur-Mer's coastal waters (France) at position 50°44.42 N, 1°34.06 W, in the eastern English Channel, and Roscoff (France) at the position 48°43.56 N, 3°58.58 W, in the western English Channel.

J. Derot et al./C. R. Geoscience xxx (2016) xxx-xxx

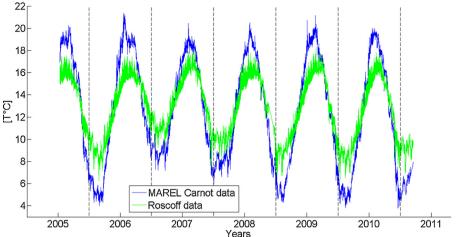


Fig. 2. Temperature MAREL Carnot (Boulogne-sur-Mer) and Roscoff data series between 2005 and 2011. Vertical lines separate each year.

Boulogne-sur-Mer (France) at position 50°44.42 N, 1°34.06 W (Fig. 1), with a periodicity of 20 min (see Zongo et al. (2011) for a presentation of this data set). Water depth at this position varies between 5 and 11 m; the measurements are taken using a floating system inserted in a tube, 1.5 m below the surface. Due to maintenance and failure of the automatic devices, there are several missing data. We consider here data from 2005 to 2011. The percentage of available data for this period is close to 88%, which corresponds to 138,574 data points acquired (Fig. 2). The MAREL Carnot station is included in the MAREL network (Automatic monitoring network for littoral environment, IFREMER, France), which is based on the deployment of moored buoys equipped with physicochemical measuring devices, working in continuous and autonomous conditions (Berthome, 1994; Blain et al., 2004; Dur et al., 2007; Schmitt et al., 2008; Woerther, 1998; Zongo and Schmitt, 2011).

2.3. Climatology

We consider here the climatology, i.e. the average interannual variations, estimated by averaging the data for each calendar day. This is shown in Fig. 3 for both temperature datasets, where the dotted lines represent upper and lower curves built using the standard deviation estimated over six different values (for a given calendar day) corresponding to the seven years of the study. The climatology of Roscoff shows a climate more maritime than in Boulognesur-Mer, which is more continental with warmer summers and colder winters. There is a minimum of around 9 °C and a maximum of around 16 °C, corresponding to an annual mean delta of 7 °C for Roscoff, while for Boulogne-sur-Mer it is close to 12 °C, with a maximum of around 18 °C and a minimum of around 6 °C. The two curves are quite smooth, showing a rather limited variability around this seasonal evolution. The standard deviation of the climatology, as

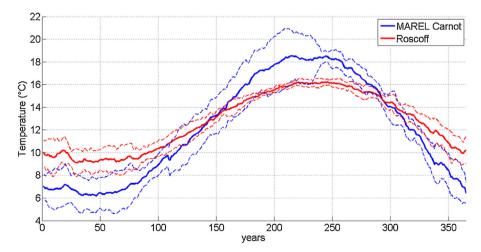


Fig. 3. The climatology of both MAREL Carnot and Roscoff temperature data: inter-annual daily averages computed using the data from 2005 to 2011. The dotted lines represent the upper and lower limits estimated using their standard deviations; the latter being estimated over seven different values, estimated yearly from 2005 to 2011. Roscoff waters are more influenced by a maritime climate: winters are warmer and summers cooler.

4

ARTICLE IN PRESS

J. Derot et al./C. R. Geoscience xxx (2016) xxx-xxx

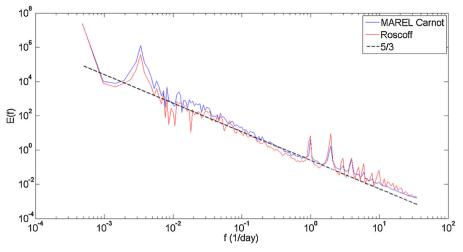


Fig. 4. Fourier power spectra of Boulogne-sur-Mer (MAREL Carnot) and Roscoff temperature time series, and a 5/3 slope straight line for comparison, corresponding to passive scalar turbulence. Some peaks are visible, corresponding mainly to seasonal, daily, and tidal cycles.

presented in Fig. 3, is more important in the Boulogne-sur-Mer area. Therefore, the inter-annual temperature variations are less marked in the Roscoff area.

3. Spectral analyses and co-spectra

We perform here spectral and co-spectral analyses. Fourier power spectra have been used in many previous studies in aquatic sciences over several decades to characterize scaling regimes (Legendre and Legendre, 2012; Platt and Denman, 1975; Seuront et al., 1996a,b; Seuront et al., 1999; Seuront et al., 2002; Winder and Cloern, 2010). To perform co-spectral analysis, a same time step is needed. Hence, for this analysis, a linear interpolation has been performed to degrade the time step to 20 min for the Roscoff dataset. Two spectra are shown in Fig. 4, in log-log plot. The spectra are clearly scaling, with slopes close to-5/3-as expected for passive scalar turbulence (Corrsin, 1951; Obukhov, 1949; Tennekes and Lumley, 1972). There are two peaks close to 12 h and 1 day, which are probably related to the impact of daily and tidal cycles. There are several slight peaks at high frequencies, and a larger peak around 90 days that could be related to the seasonal cycle.

3.1. Co-spectral analyses

In order to compare the two dynamics at different scales, we have estimated the co-spectrum, defined as (Bendat and Piersol, 2000):

$$R_{T_1,T_2}(f) = \frac{\left| E_{T_1,T_2}(f) \right|}{\sqrt{E_{T_1}(f)E_{T_2}(f)}} \tag{1}$$

where $E_{XY}(f)$ is the co-spectrum of X and Y. When there is a linear relation between X and Y, R_{XY} is equal to 1, and for uncorrelated processes $R_{XY} = 0$ (Bendat and Piersol, 2000; Lumley, 2007). We find that there is a good correlation for scales larger than 172 days; there is a transition between 11 days and 172 days, and a decorrelation for smaller

scales (Fig. 5). The average residual velocity of the surface currents (after subtraction of tidal currents) in the English Channel was measured using very high frequency radars as being between 0.4 and 0.45 m/s (Sentchev and Yaremchuk, 2007), and the distance between the two stations is around 460 km. Therefore, we can assume a decorrelation time of $460 \ 10^3/0.425 \approx 12.5$ days. This is close to the transition scale of about 11 days, which is found in Fig. 5. Hence, we may say that high-frequency (scale < 11 days) temperature fluctuations are independent, whereas lower frequencies are connected.

4. Time-dependent intrinsic correlation analysis (TDIC)

4.1. TDIC and EMD

We perform here a time-frequency analysis using Empirical Mode Decomposition (EMD), which was first introduced at the end of the 1990s in the field of physical oceanography (Huang et al., 1998, 1999). It has since been very successful with applications in many fields of science, and has received several thousand citations. The aim of this method is to decompose a time series in a sum of time series called "modes", each having a characteristic frequency. At the end of the decomposition, the method expresses the analyzed signal X(t) as a sum of mode time series $C_i(t)$ and a residual r(t):

$$X(t) = \sum_{i=1}^{n} C_i(t) + r(t)$$
(2)

The mode selection is adaptive: this means that the mode shape $C_i(t)$ is not decided *a priori*, but estimated from the data using an algorithm based on successive interpolation passing through local maxima. The precise algorithm is not repeated here. We refer to Huang et al. (1998) or Huang et al. (2008, 2009) for details. A new correlation method, based on EMD, has recently been introduced (Chen et al., 2010). The method uses EMD to decompose two time series into modes $C_{1,i}$ and $C_{2,i}$ (i = 1, ..., n) and consider the

J. Derot et al./C. R. Geoscience xxx (2016) xxx-xxx



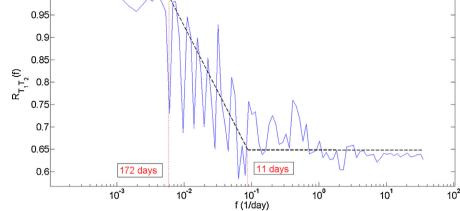


Fig. 5. Co-spectra in semi-log (x axis) plot, of the temperature series in Roscoff and Boulogne-sur-Mer. There is a good correlation for scales larger than 172 days, a transition between 172 and 11 days, and a relative decorrelation for scales smaller than 11 days.

correlation between them. The main idea is to consider that the classical correlation implicitly assumes stationarity and linearity of both time series. The proposed method is thus a generalization. A local correlation is then estimated between two modes having a similar characteristic scale. Since the method is designed to work also for non-stationary series, its estimation is made locally, as well as globally, for the whole series. The EMD method, together with the Hilbert spectral analysis, is able to extract, for each mode, a local frequency. Each mode has a characteristic frequency, corresponding to a given time scale T_{i} . There are thus the time scales $T_{1,i}$ and $T_{2,i}$ for the mode *i* of series No. 1 and No. 2. The correlation is estimated over a local interval centered around *t*:

$$I_{t_w}(t) = \left[t - \frac{t_w}{2}; t + \frac{t_w}{2}\right]$$
(3)

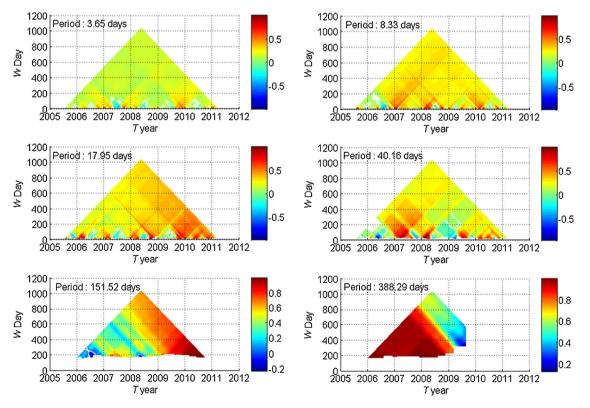


Fig. 6. Cross-correlation plot between Roscoff and Boulogne-sur-Mer temperature series, using the TDIC method, based on empirical mode decomposition. The correlation is done here for modes having similar scales. The cross-correlation is time dependent (horizontal axis), and estimated over interval of increasing widths (vertical axis). Color scales emphasize strong events of correlation or anti-correlation.

where t_w is the width of the interval: $t_w \ge t_d = \max(T_{1,i}, T_{2,i})$. The local correlation writes (Chen et al., 2010):

$$R_{t_{w}}(t) = corr \ \left\{ I_{\{t_{w}\}} \right\} \left(C_{1,i}(t) C_{2,i}(t) \right) \tag{4}$$

where *corr* is the usual correlation coefficient, *t* is time, and *corr{l_t}* symbolizes a correlation estimated over the interval $t \in I$. This is performed for each mode i: $R_{t_w}(t)$ is estimated for each *t* value (horizontal axis), for intervals of width $R_{t_w}(t)$ (vertical axis). A color code is used for the value of the correlation. The usual correlation coefficient is the value obtained at the top of the triangle, when all values have been used to estimate the local correlation. This powerful method was introduced in 2010 (Chen et al., 2010) and has since met growing interest, and has been used in several recent studies (Hong and Wang, 2014; Huang and Schmitt, 2014; Yin et al., 2014).

Here to perform the TDIC method (Fig. 6), we used a Matlab code written by Yongxiang Huang (Huang and Schmitt, 2014); this code is itself based on the code written by Chen and Huang (Chen et al., 2010). The TDIC method has the advantage of giving a correlation for each characteristic scale, but for our time series with a time step of 20 min, it was not possible to enter directly the raw data, because of memory limitation to perform this computation. In this context, we have degraded our time series to perform the experiment reported in Fig. 6, where the new time step used is 1 day. This is not affecting information on time scales larger than one day, which are the concern here.

4.2. Application of the TDIC method

Fig. 6 represents six TDIC graphs, with a characteristic time between 3.65 and 388.00 days. A pattern is visible on the first four TDIC graphs (until 40 days). For each year, on the baseline there is a blue patch in summer and a red patch in winter, which means that there is a correlation in winter and an anti-correlation in summer. Considering larger time lags, at the scale of 3 days (first graph) there seems to be no correlation. On the contrary, for scales of 8, 18, and 40 days (graphs 2, 3, and 4), the color scale images show global correlations between the two series. For years 2006-2008, the correlation is stronger at the scale of 40 days, whereas for scales 2009-2011 the correlation is stronger at the scale of 18 days. On the last graph with a scale of 388 days, there is a large red correlation patch on the left side and a blue no-correlation patch on the right side. This pattern could be because the amplitudes of the temperature signal are very close for years 2007 and 2008.

5. Discussion and conclusion

We have performed a comparison between two automated fixed-point systems located in the English Channel. Firstly, we have considered the climatology for both areas, and we have seen that Boulogne-sur-Mer was situated in a more temperate climate zone than Roscoff. Secondly, we have observed that both temperature spectral slopes were close to 5/3, which is the spectral slope of a passive scalar in fully developed turbulence. Finally, both correlation methods have converged toward a transition scale of 11 days for both series: scales smaller than 11 days are not correlated and larger scales possess some correlation.

The Roscoff system is installed on a bottom mooring, contrary to MAREL Carnot, which records data from the surface. Both sites are in coastal areas with strong tidal currents and consequently the water mass is thoroughly mixed. The spectral analysis in section 3 reveals that mean forcings, which seem to have an influence on temperature, are cycles linked to day/night, tidal, seasons, and high tides. These forcings are mainly controlled by astronomical mechanisms. Overall, the surface currents in the English Channel go east toward the strait, from the Atlantic to the North Sea. The surface temperatures from Roscoff could influence the surface temperature in Boulogne-sur-Mer, with a characteristic time of 11 days corresponding to an average transport time. From the results of co-spectral analysis, a link between both series was found for scales larger than 11 days, which is also the time scale found by considering a mean velocity of 0.5 m/s and a distance of 460 km. The TDIC analysis provided results consistent with this time scale. This new method is also useful to locally express correlation events. Here, we found that for scales from 8 to 40 days, there is a correlation in winters and an anti-correlation is summers.

Fixed-point systems provide temporal information. We can draw a parallel with the hypothesis proposed by Taylor in turbulence (Taylor, 1935), which allows the transformation of temporal information in the turbulence field on spatial information, when we know the mean velocity. In this context, we could compare the events recorded by automatic systems to the outputs of a numerical model. This could give us information to refine the model output. With an average current speed of 0.5 m/s in our study area and a sampling frequency of 20 min, we have a distance below 1 km: this scale is consistent with the spatial resolution of currently used models (Beaman et al., 2011; Lim et al., 2013; Wolanski and Kingsford, 2014). Another perspective is to consider a possible link with the NAO (North Atlantic Oscillation) variability, as performed in a recent study, where the authors considered the link between sea surface temperature and the NAO (Tréguer et al., 2014).

Acknowledgements

The region "Nord–Pas-de-Calais" (France) and the "Agence de l'eau Artois Picardie" are thanked for the financial support granted to Jonathan Derot's PhD thesis. IFREMER is acknowledged for access to MAREL data and Thierry Cariou for access to Estacade data; we thank Alain Lefebvre and Michel Repecaud from IFREMER for discussions. Yongxiang Huang is thanked for discussions on the TDIC method. We thank Denis Marin (LOG) for realizing Fig. 1. The TDIC code is free to access on the following website: http://www.fg-schmitt.fr/EMD_and_ multifractals.html. www.englisheditor.webs.com is also thanked for its English proofing.

J. Derot et al./C. R. Geoscience xxx (2016) xxx-xxx

References

- Beaman, R.J., O'Brien, P.E., Post, A.L., De Santis, L., 2011. A new highresolution bathymetry model for the Terre Adélie and George V continental margin. East Antarctica. Antarct. Sci. 23 (01), 95–103.
- Bendat, J.S., Piersol, A.G., 2000. Random data: analysis and measurement procedures. John Wiley & Sons.
- Berthome, J., 1994. MAREL : un réseau automatisé de veille pour l'environnement littoral. Equinoxe 34–35.
- Blain, S., Guillou, J., Tréguer, P., Woerther, P., Delauney, L., Follenfant, E., Gontier, O., Hamon, M., Leilde, B., Masson, A., 2004. High frequency monitoring of the coastal marine environment using the MAREL buoy. J. Environ. Monit. 6, 569–575.
- Chang, G., Dickey, T., 2001. Optical and physical variability on timescales from minutes to the seasonal cycle on the New England shelf: July 1996 to June 1997. J. Geophys. Res.: Oceans 106, 9435–9453.
- Chavez, F.P., Pennington, J.T., Herlien, R., Jannasch, H., Thurmond, G., Friederich, G.E., 1997. Moorings and drifters for real-time interdisciplinary oceanography. J. Atmos. Ocean. Tech. 14, 1199–1211.
- Chen, X., Wu, Z., Huang, N.E., 2010. The time-dependent intrinsic correlation based on the empirical mode decomposition. AADA 2, 233–265.
- Corrsin, S., 1951. On the spectrum of isotropic temperature fluctuations in an isotropic turbulence. J. Appl. Phys. 22, 469–473.
- Dickey, T.D., 1991. The emergence of concurrent high-resolution physical and bio-optical measurements in the upper ocean and their applications. Rev. Geophys. 29, 383–413.
- Dickey, T.D., Douglass, R., Manov, D., Bogucki, D., Walker, P., Petrelis, P., 1993. An experiment in two-way communication with a multivariable moored system in coastal waters. J. Atmos. Ocean. Tech. 10, 637–644.
- Dur, G., Schmitt, F., Souissi, S., 2007. Analysis of high frequency temperature time series in the Seine estuary from the Marel autonomous monitoring buoy. Hydrobiologia 588, 59–68.
- Hong, W., Wang, J., 2014. Multiscale behavior of financial time series model from Potts dynamic system. Nonlinear Dynam. 78 (2), 1065–1077.
- Huang, Y., Schmitt, F.G., Lu, Z., Liu, Y., 2008. An amplitude-frequency study of turbulent scaling intermittency using Hilbert spectral analysis. EPL 84, 40010.
- Huang, N.E., Shen, Z., Long, S.R., 1999. A new view of nonlinear water waves: The Hilbert Spectrum 1. Annu. Rev. Fluid. Mech. 31, 417–457.
- Huang, N.E., Shen, Z., Long, S.R., Wu, M.C., Shih, H.H., Zheng, Q., Yen, N.-C., Tung, C.C., Liu, H.H., 1998. The empirical mode decomposition and the Hilbert spectrum for nonlinear and non-stationary time series analysis. Proc. Roy. Soc. London. Ser. A: Math. Phys. Eng. Sci. 454, 903–995.
- Huang, Y., Schmitt, F.G., 2014. Time dependent intrinsic correlation analysis of temperature and dissolved oxygen time series using empirical mode decomposition. J. Mar. Syst. 130, 90–100.
- Huang, Y., Schmitt, F.G., Lu, Z., Liu, Y., 2009. Analysis of daily river flow fluctuations using empirical mode decomposition and arbitrary order Hilbert spectral analysis. J. Hydrol. 373, 103–111.
- Legendre, P., Legendre, L.F., 2012. Numerical ecology, Elsevier, 853 p. Lim, H.S., Chun, I., Kim, C.S., Park, K.S., Shim, J.S., Yoon, J.J., 2013. High-
- Lim, H.S., Chun, I., Kim, C.S., Park, K.S., Shim, J.S., Yoon, J.J., 2013. Highresolution operational coastal modeling system for the prediction of hydrodynamics in Korea using a wavecurrent coupled model. J. Coast. Res. 314–319.

Lumley, J.L., 2007. Stochastic tools in turbulence. Courier Dover Publications.

- Nam, S.H., Kim, G., Kim, K.R., Kim, K., Oh, L., Kim, K.W., Ossi, H., Kim, Y.G., 2005. Application of real-time monitoring buoy systems for physical and biogeochemical parameters in the coastal ocean around the Korean peninsula. J. Mar. Sci. Technol. 39, 54–64.
- Obukhov, A.M., 1949. The structure of the temperature field in a turbulent flow. Izv. Akad. Nauk. SSSR, Ser. Geogr. and Geophys. 13, 58–69.
- Platt, T., Denman, K.L., 1975. Spectral analysis in ecology. Annu. Rev. Ecol. Syst. 6, 189–210.
- Schmitt, F.G., Dur, G., Souissi, S., Brizard Zongo, S., 2008. Statistical properties of turbidity, oxygen and pH fluctuations in the Seine river estuary (France). Physica A 387, 6613–6623.
- Sentchev, A., Yaremchuk, M., 2007. VHF radar observations of surface currents off the northern Opal coast in the eastern English Channel. Cont. Shelf Res. 27, 2449–2464.
- Seuront, L., Gentilhomme, V., Lagadeuc, Y., 2002. Small-scale nutrient patches in tidally mixed coastal waters. Mar. Ecol. Prog. Ser. 232, 29–44.
- Seuront, L., Schmitt, F., Lagadeuc, Y., Schertzer, D., Lovejoy, S., 1999. Universal multifractal analysis as a tool to characterize multiscale intermittent patterns: example of phytoplankton distribution in turbulent coastal waters. J. Plankton. Res. 21, 877–922.
- Seuront, L., Schmitt, F., Lagadeuc, Y., Schertzer, D., Lovejoy, S., Frontier, S., 1996a. Multifractal analysis of phytoplankton biomass and temperature in the ocean. Geophys. Res. Lett. 23, 3591–3594.
- Seuront, L., Schmitt, F., Schertzer, D., Lagadeuc, Y., Lovejoy, S., 1996b. Multifractal intermittency of Eulerian and Lagrangian turbulence of ocean temperature and plankton fields. Nonlin. Processes Geophys. 3, 236–246.
- Taylor, G.I., 1935. Statistical theory of turbulence. Proc. Roy. Soc. Lond. Ser. A 151, 421–444.
- Tennekes, H., Lumley, J.L., 1972. A first course in turbulence. MIT press.
- Tréguer, P., Goberville, E., Barrier, N., L'Helguen, S., Morin, P., Bozec, Y., Rimmelin-Maury, P., Czamanski, M., Grossteffan, E., Cariou, T., 2014. Large and local-scale influences on physical and chemical characteristics of coastal waters of western Europe during winter. J. Mar. Syst. 139, 79–90.
- Winder, M., Cloern, J.E., 2010. The annual cycles of phytoplankton biomass. Phil. Trans. R. Soc. B. 365, 3215–3226.
- Woerther, P., 1998. Marel : Mesures automatisées en réseau pour l'environnement littoral. L'Eau, l'industrie, les nuisances 67–72.
- Wolanski, E., Kingsford, M.J., 2014. Oceanographic and behavioural assumptions in models of the fate of coral and coral reef fish larvae. J. R. Soc. Interface 11 (98), 20140209.
- Yin, L., Qiao, F., Zheng, Q., 2014. Coastal-Trapped Waves in the East China Sea Observed by a Mooring Array in Winter 2006. J. Phys. Oceanogr. 44 (2), 576–590.
- Zongo, S., Schmitt, F.G., 2011. Scaling properties of pH fluctuations in coastal waters of the English Channel: pH as a turbulent active scalar. Nonlin. Processes Geophys. 18, 829–839.
- Zongo, S., Schmitt, F.G., Lefebvre, A., Répécaud, M., 2011. Observations biogéochimiques des eaux côtières à Boulogne-sur-Mer à haute fréquence: les measures automatiques de la bouée MAREL. In: Schmitt, F.G. (Ed.), Observation des écosystèmes marins et terrestres de la Côte d'Opale : du naturalisme à l'écologie. Presses de l'UOF, Paris, pp. 253–266.