1 Supplementary Material A: Summary of field data used in the study.

- 2 3 Table SMA - Characteristics (source, LON (longitude), LAT (latitude), time frame, and sampling frequency) of
- the field sampling sites.

Source – Variable (Station)	LON [°]	LAT [°]	Time frame	Sampling frequency
MAGEST – Turbidity (Le Verdon)	-1.0421	45.5438	2017 - 2021	10-30 min
MAGEST – Turbidity (Pauillac)	-0.7426	45.1985	2017 - 2018	10-30 min
GEMMES – Turbidity (buoy 20)	-1.3917	45.6375	2016/11 - 2017/09	15 min
HydroEAU – River discharge (Garonne, Dordogne)			1959 - 2021	daily
Meteo France – Wind (Cap Ferret)	-1.2480	44.3160	1985 - 2022	3-hourly
Meteo France – Wind (Royan)	-0.9683	45.6317	1991 - 2022	hourly
Meteo France – Wind (Pauillac)	-0.7828	45.2142	2004 - 2020	hourly
Meteo France – Wind (Merignac/Bordeaux)	-0.6900	44.8300	2004 - 2020	hourly

5 Supplementary Material B: Comparativeness of satellite derived turbidity and field stations

6 Match-ups between Sentinel-3/OLCI and field turbidity data were identified originating from satellite

7 overpasses quality-controlled for flags caused by clouds or other atmospheric interferences. Water reflectance

8 of regions of satellite scenes with flags indicating such artifacts were removed and not considered for match-9 ups. From these match-ups, an empirical algorithm was established aided by a Support Vector Machine (SVM)

10 model. The comparison between turbidity estimates using the red and near infra-red bands (Figure SMA)

11 resulted in good agreement with field turbidity: $r_{kendal-t} = 0.72$, p-value << 0.0001; n = 174.



12

13 Figure SMB – Type-II linear regression (black line) between field turbidity and Sentinel-3 derived turbidity. 14 Blue-triangle symbols represent field turbidity from Pauillac station, Light-gray symbols represent field 15 turbidity from Le Verdon station and Yellow-squared symbol represents field data from the GEMMES network in buoy 20. Dashed grey line is 1:1.

17 Supplementary Material C: Comparativeness and bias correction of wind from ERA5 (ECMWF)

18 To overcome systematic errors from reanalysis ECMWF ERA-5 wind speeds, we performed a bias correction 19 Cumulative Distribution Function bv applying the (CDF) matching approach 20 (https://nl.mathworks.com/matlabcentral/fileexchange/78784-cdf-matching-bias-correction-method-in-matlab; 21 accessed on August, 2023, by Singh et al., 2020 and Reichle & Koster, 2004) for the years between 2005 and 2021. This is one of the widely used statistical methods to minimise the bias in a derived or modelled data 22 23 compared with a more reliable source of data (usually field). We adjusted the ECMWF wind speed (u and v 24 components) according to the CDF of field wind speeds to minimise their difference. The data was fit to a 25 polynomial function of 3^{rd} order (y = p₁(x³) + p₂(x²) + p₃(x¹) + p₄), then bias corrected with the coefficients

26 shown in Table SM3.

27 Table SMC – Coefficients estimated from CDF matching for BIAS correctio	n.
----------------------------------------------------------------------------	----

Wind component	\mathbf{p}_1	\mathbf{p}_2	\mathbf{p}_3	p_4
u	-0.000367	0.0153	-0.1101	-0.1366
V	-0.001600	0.0049	-0.0302	0.2107

Figure SMC-a,b shows the reference (field data), the biased and the bias-corrected wind speed distribution. 28





30 Figure SMC - Bias correction using the CDF matching technique. Solid black line is the CDF of field measured wind speed for u (a) and v (b) wind speed components, dashed and dotted lines are the CDF of 31 biased and bias-corrected wind speed from reanalysis ERA-5 ECMWF data, respectively

33 Supplementary Material D: Maximum wind speeds

- 34 Spatial distribution of maximum winds speeds from ECMWF-ERA5 bias corrected wind speeds, and field wind
- 35 speeds.



- 37 Figure SMD Maximum wind speeds of the time series of bias corrected ECMWF ERA5 (2016 to 2021) and
- 38 field data at stations Cap Ferret, Merignac/Bordeaux, Pauillac and Royan with timeseries constrained between
- 39 years 2016 and 2021. Of the four field stations, stations Pauillac and Merignac/Bordeaux extend until 2020.

40 Supplementary Material E: Statistic metrics of non-parametric tests Kruskal-Wallis and Dunn-Sidak

41 The non-parametric test Kruskal-Wallis H tests the null hypothesis (H₀) that the data in each column

42 of a matrix comes from the same distribution. The alternative hypothesis is that not all samples come

43 from the same distribution. For database A and database B p-values were below 0.0001 for all hydro-

44 sedimentological variables.

45 Building upon this, the Dunn-Sidak test was employed for pairwise comparisons to identify which specific groups of classes were statistically different. In the test, the null hypothesis (H₀) states that 46 there is no difference between the ranks of the two classes being compared. If the O-value exceeds the 47 48 critical value, we reject H₀, indicating a statistically significant difference between the two classes. 49 Conversely, if the *O-value* is below the critical value, we fail to reject H₀, suggesting no significant difference between pair of hydro-sedimentological classes. Therefore, rejecting H₀ implies that the 50 51 group being compared is statistically different from the other. If at least one variable (either river 52 discharge, tidal current, water level, wind speed, depth or suspended sediment) for a given class is 53 statistically significant different from the other classes, then that class is different from the remaining 54 classes. Table SME-1 and Table SME-2, below, show the Q-values from the Dunn-Sidak tests for

55 database A and database B, respectively.

56 Table SME-1 - Dunn-Sidak test results for Database A. The table presents the Q-values from pairwise

57 group comparisons. Comparisons with *Q*-values in **bold** indicate that the null hypothesis cannot be

rejected (Q-value < critical value of 3.25), suggesting no significant difference between the respective 58

59 classes for a certain variable (river discharge, tidal current, water level, wind speed, depth or suspended

60 sediment).

	/										
		C6	C3	C4	C7	C1	C5	C2	C8	C9	C10
	C6										
	C3	83.2									
e	C4	50.3	65.8								
larg	C7	21026.0	15800.0	2991.8							
scł	C1	117.5	5.9	67.4	20711.0						
r di	C5	148.7	20.9	70.3	22547.0	20.1					
ive	C2	388.1	201.0	104.5	21889.0	259.4	262.9				
R	C8	248.3	271.7	10.2	18388.0	348.4	390.5	593.9			
	C9	58.4	44.2	58.2	22491.0	68.2	97.7	356.3	312.1		
	C10	166.4	205.4	10.3	12488.0	236.2	255.1	395.8	1.2	203.6	
	C6										
	C3	364.9									
t	C4	40.1	109.4								
ren	C7	12241.0	9511.9	1730.9							
cur	C1	94.9	289.4	53.9	12083.0						
al e	C5	219.7	219.5	69.6	13257.0	113.4					
Tid	C2	565.9	46.5	119.1	13078.0	456.4	380.4				
	C8	175.8	202.0	68.3	11022.0	88.7	6.4	318.4			
	C9	233.8	209.8	71.4	13290.0	127.1	14.9	366.8	18.4		
	C10	95.1	174.4	61.8	7460.9	36.9	28.3	236.7	22.5	36.2	

63	Table SME-1 -	(continued)
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		C6	C3	C4	C7	C1	C5	C2	C8	C9	C10
	C6										
	C3	490.1									
	C4	57.6	36.3								
vel	C7	11471.0	9061.3	1717.2							
e le	C1	573.4	54.3	26.1	11802.0						
ater	C5	407.7	213.6	3.0	12623.0	212.3					
W,	C2	847.6	126.8	60.2	12569.0	242.7	489.0				
	C8	451.9	102.5	15.4	10613.0	64.6	120.2	285.0			
	C9	513.7	137.9	11.1	12745.0	109.8	113.5	381.7	28.7		
	C10	297.3	84.8	13.4	7199.3	51.7	70.9	198.4	7.4	10.9	
	C6										
	C3	112.3									
	C4	83.7	104.5								
eed	C7	11741.0	8888.2	1615.0							
sb	C1	188.9	30.7	111.3	11687.0						
ind	C5	92.6	49.6	96.3	12600.0	110.6					
Ň	C2	377.9	164.1	136.6	12382.0	176.5	309.6				
	C8	102.3	24.4	100.0	10507.0	66.9	27.2	227.9			
	C9	275.9	315.2	47.0	12256.0	470.6	397.4	688.7	347.9		
	C10	157.6	218.3	45.2	6009.1	270.6	214.0	380.9	216.2	3.7	
	C6										
	C3	122.8									
	C4	135.0	157.3								
_	C7	1654.2	12487.0	2258.4							
ptł	C1	182.6	132,2	161.6	16383.0						
Ď	C5	210.4	278.8	107.0	17416.0	399.7					
	C2	326.1	115.2	180.6	17242.0	132.2	565.3				
	C8	4.9	118.0	133.8	14655.0	166.9	178.0	291.6			
	C9	141.9	229.4	116.2	17510.0	333.1	74.2	495.5	118.3		
	C10	88.8	165.1	111.7	9866.6	198.6	29.4	281.2	81.5	9.9	
	C6										
nt	C3	38.1									
me	C4	29.2	36.2								
edi	C7	10359.0	7780.9	1469.6							
d s	C1	102.2	42.1	44.6	10251.0						
Jude	C5	385.9	238.8	80.9	11416.0	263.7					
ipei	C2	530.2	351.3	103.2	11118.0	409.9	171.5				
Sus	C8	133.7	69.2	50.6	9312.6	37.2	195.4	329.9			
V 1	C9	242.4	135.2	61.6	11291.0	123.4	155.4	319.8	70.2		
1	C10	168.8	121.8	68.3	6401.9	103.3	47.9	141.4	74.9	34.3	

Running Title

65 **Table SME-2** - Dunn-Sidak test results for Database B. The table presents the *Q*-values from pairwise

66 group comparisons. Comparisons with *Q*-values in **bold** indicate that the null hypothesis cannot be

67 rejected (*Q-value* < critical value of 3.25), suggesting no significant difference between the respective 68 classes for a certain variable (river discharge, tidal current, water level, wind speed, depth or suspended

69 sediment). The symbol '-' indicates that the Dunn-Sidak test accepts the null hypothesis (*Q*-value \leq 70 critical value)

		C6	C3	C4	C7	C1	C5	C2	C8	С9	C10
	C6										
	C3	345.9									
ge	C4	207.3	554.8								
nar	C7	50.4	286.1	252.7							
iscl	C1	36.2	395.0	178.2	87.4						
r d	C5	187.1	541.0	23.9	233.7	156.8					
ive	C2	43.8	353.6	143.0	88.4	12.6	123.5				
R	C8	178.1	-	300.5	146.4	200.9	288.4	198.7			
	C9	25.2	202.4	164.3	9.2	49.3	149.6	54.9	130.2		
	C10	13.3	1.8	22.6	10.9	14.8	21.5	15.2	0.2	11.6	
	C6										
	C3	479.6									
It	C4	39.0	514.8								
rer	C7	191.9	271.9	228.3							
cur	C1	452.2	32.2	487.9	243.5						
lal	C5	350.8	124.7	386.8	149.2	94.1					
Tid	C2	267.5	155.6	300.6	89.7	128.7	44.8				
	C8	163.4	113.7	186.0	47.2	95.9	41.8	11.9			
	C9	156.1	158.2	181.6	24.8	138.2	76.3	40.8	21.3		
	C10	-	19.7	3.0	7.6	18.4	14.3	12.1	11.2	24.8	
	C6										
	C3	506.9									
	C4	154.6	343.5								
ve]	C7	-	491.5	147.9							
r le	C1	299.2	217.4	136.5	288.6						
ate	C5	361.0	141.7	201.5	349.4	71.0					
\mathbb{A}	C2	281.9	165.3	140.7	273.8	22.5	39.4				
	C8	114.9	178.4	22.7	112.0	56.6	96.4	67.6			
	C9	173.0	159.1	68.4	169.0	20.8	66.2	35.0	31.7		
	C10	0.6	22.7	7.6	0.8	13.6	16.6	14.7	9.3	12.2	
	C6										
	C3	127.9									
-	C4	158.9	289.4								
eed	C7	94.0	28.7	248.6							
sb	C1	4.8	137.2	159.4	101.6						
ind	C5	62.3	193.0	98.7	155.7	59.6					
M	C2	253.2	143.9	394.2	163.9	164.1	310.9				
	C8	35.0	109.6	59.5	91.0	32.6	1.4	193.5			
	C9	238.1	156.4	343.8	172.1	244.6	280.5	47.2	205.0		
1	C10	15.6	10.0	22.8	11.3	15.8	18.3	3.0	18.2	0.3	

•10			(commuee	.)								
			C6	C3	C4	C7	C1	C5	C2	C8	С9	C10
		C6										
		C3	356.2									
		C4	71.6	425.9								
	_	C7	357.5	18.1	424.2							
	pth	C1	40.8	410.4	33.5	408.5						
	De	C5	304.3	47.7	373.5	63.4	355.7					
		C2	319.3	7.4	381.1	8.8	364.0	49.1				
		C8	123.9	81.9	166.0	91.1	148.8	54.1	82.6			
		C9	206.6	25.8	253.7	37.3	236.3	5.3	29.4	49.0		
		C10	2.7	12.8	6.0	13.6	4.5	10.7	319.3	13.2	11.1	
		C6										
	nt	C3	169.7									
	neı	C4	225.0	62.4								
	ribe	C7	255.3	95.6	33.0							
	d se	C1	19.3	155.8	213.2	244.5						
	qee	C5	335.2	172.3	105.0	70.0	327.0					
	pen	C2	277.5	131.4	73.1	42.3	267.6	20.7				
	lsng	C8	67.1	30.9	66.7	86.3	56.7	129.3	110.8			
	01	C9	167.4	57.4	15.6	7.0	157.1	54.7	37.2	67.5		
		C10	5.8	13.2	16.0	17.5	6.6	20.6	19.6	10.9	17.0	

Table SME-2 - (continued)

74 Supplementary Material F: Performance metrics of Random Forest algorithms by database and hydro-

75 sedimentological class.

me	diles of Rai		i algoritim	lis ill database 11 per ela.	
		Class	\mathbf{R}^2	RMSE log ₁₀ [g.m ⁻³]	
		C6	0.93	0.0941	
	ns Br	C3	0.92	0.1104	
	tio	C4	0.91	0.1281	
	eva ndi	C7	0.91	0.1116	
	Pro	C5	0.91	0.1424	
		C2	0.92	0.1331	
	me ts	C8	0.94	0.1537	
	tre /en	C9	0.94	0.1221	
	Ex	C10	0.93	0.1425	

76 Table SMF-1 – Performance metrics of Random Forest algorithms in database A per class.

77

78 Table SMF-2 – Performance metrics of Random Forest algorithms in database B per class.

	Class	\mathbf{R}^2	RMSE log ₁₀ [g.m ⁻³]
	C6	0.94	0.0635
lg US	C3	0.92	0.2054
tio	C4	0.95	0.0848
eva ndi	C7	0.96	0.0697
Pr_{col}	C5	0.92	0.1839
	C2	0.95	0.0979
ne ts	C8	0.98	0.1222
trer	C9	0.96	0.0940
Ext ev	C10	0.88	0.1013

80 Supplementary Material G: Satellite-based relative contribution to variability of turbidity (database B)

81 Relative contribution results using database B and the proposed hydro-sedimentary framework. Figure SMG-1

82 depicts spatial variability of relative contribution of forcings to turbidity, Figure SMG-2 shows the distributions

of the 10 identified hydro-sedimentological classes, and Figure SMG-3 depict the class-specific relative 83

84 contribution of forcings.



85 86

Figure SMG-1 - Synoptical mean contribution of predicting factors for satellite-derived turbidity (database B, 87 with satellite-based turbidity): (a) river discharge, (b) tides, (c) winds and (d) uncertainties/noise. Attention to 88 the different ranges of percentage on each subplot.



89 90

90Figure SMG-2- Classes of coherent hydrodynamical and sedimentological patterns of database B (x-axis), for91river discharge (a,g, log_{10} transformed), tidal current (b,h), water level (c,i), wind speed (c,j), depth (e,k) and92satellite-derived turbidity (Turb log_{10} transformed; f,l). Left panels (a-f) represent hydro-sedimentological93classes under prevailing conditions, while right panels (g-i) represent classes under extreme event conditions94(respectively simultaneous extreme river discharge and potential windburst, extreme river discharge and95potential windburst). The shaded upper portion of (a,g) and (d,j) respectively represent the threshold defining96occurrence of extreme river discharge ($Q_{log10} > 3.24 \text{ m}^3.\text{s}^{-1}$) and potential windburst events ($U > 6.82 \text{ m.s}^{-1}$) in97the Gironde Estuary. Classes are ordered based on increasing turbidity.

Running Title



Pierr Discharge Tide Wind Noise
Figure SMG-3 - Class-specific relative contribution (%) of predicting factors (river discharge, tide, wind) for
satellite-derived turbidity (database B) and identified uncertainty or noise. Classes C1-C7 depict relative
contribution of forcing mechanisms under prevailing conditions (ordered in classes from low (C6) to high (C1)
log SSC), classes C8-C10 depict relative contribution of extreme events (extreme high river discharge,

103

log SSC), classes C8-C10 depict relative contribution of extreme events (extreme high river discharge, potential windbursts, and simultaneous extremes, respectively).