

# Evaluation of the sampling uncertainty from the Monte Carlo Simulation of georeferenced information

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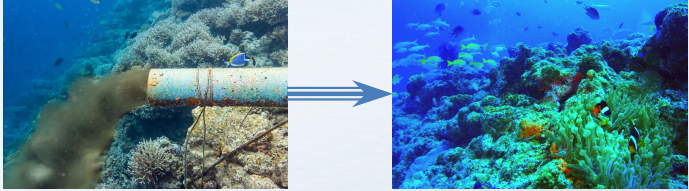
## Overview

- Introduction
- Assessment of pollution trends
- Transitional vs oceanic waters
- Sampling Uncertainty Modelling
- Application examples:
  - Phosphate levels in the Tagus Estuary
  - Phosphate levels in oceanic coastal waters
- Conclusions




**Introduction**

- Environmental protection aims to:
  - preserve the existing natural environment
  - to repair damage and reverse trends, when possible
- How?
  - Identifying and characterizing pollution sources
  - Limiting or eliminating its effects



- Monitoring plays a major role in environmental protection policies
- The assessment of trends is essential, which requires long term monitoring (e.g. ocean acidification or eutrophication studies)

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**Assessment of pollution trends**

**The assessment of pollution trends is affected by environment heterogeneity.**

The mean value of a component in a large environmental area is robust to system heterogeneity (...)

The modelling of system heterogeneity depends on:




- » Number and positions from where samples are collected;
- » Determined concentration values that express concentration profiles;
- » Uncertainty of position coordinates;
- » Uncertainty of sample analysis;

*(defines system models uncertainty)*

Evaluated MU depends on available information and the detail of their modelling.

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## Transitional vs oceanic waters

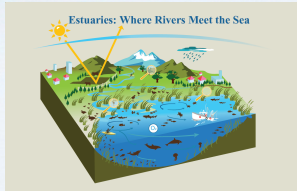
A comparison

### Estuarine Water

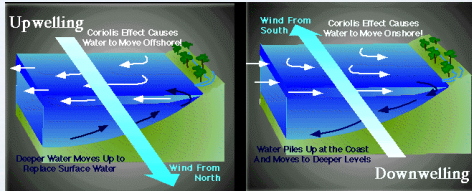
- Seasonality – freshwater flow and temperature
- Low scale spatial gradients
- Anthropogenic activities
- Tidal effects

### Ocean Water

- Seasonality – currents and wind
- Large scale spatial gradients
- Nutrient poor/depleted superficial water layers






Estuaries: Where Rivers Meet the Sea



Upwelling: Coriolis Effect Causes Water to Move Offshore. Wind From North. Deeper Water Moves Up to Replace Surface Water.

Downwelling: Wind From South. Coriolis Effect Causes Water to Move Onshore. Water Piles Up at the Coast and Moves to Deeper Levels.

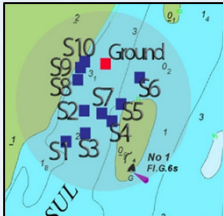
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## Sampling Uncertainty Modelling

### Methodology:

- 1) Collect samples at a defined depth and known GPS coordinates

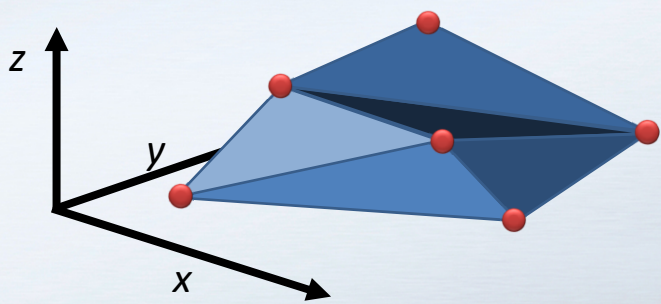


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**Sampling Uncertainty Modelling**

**Methodology:**

- 1) Collect samples at a defined depth and known GPS coordinates
- 2) Describe collected information as a 3D surface  $(x, y, z)$ , where  $x$  and  $y$  are GPS coordinates and  $z$  samples concentrations

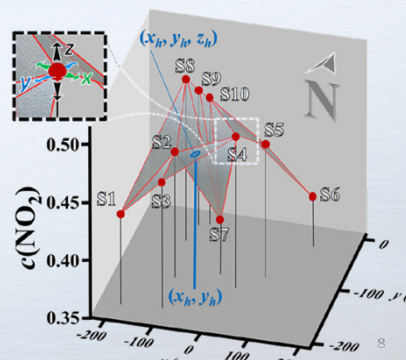


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**Sampling Uncertainty Modelling**

**Methodology:**

- 1) Collect samples at a defined depth and known GPS coordinates
- 2) Describe collected information as a 3D surface  $(x, y, z)$ , where  $x$  and  $y$  are GPS coordinates and  $z$  samples concentrations
- 3) Randomization of the 3D surface given GPS coordinates and sample analysis precision



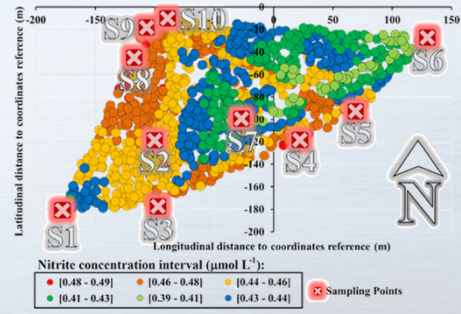









### Sampling Uncertainty Modelling

**Methodology:**

- 1) Collect samples at a defined depth and known GPS coordinates
- 2) Describe collected information as a 3D surface (x, x, z), where x and y are GPS coordinates and z samples concentrations
- 3) Randomization of the 3D surface given GPS coordinates and sample analysis precision
- 4) Random scanning of the randomised surfaces to model concentration variation (Single Sampling)

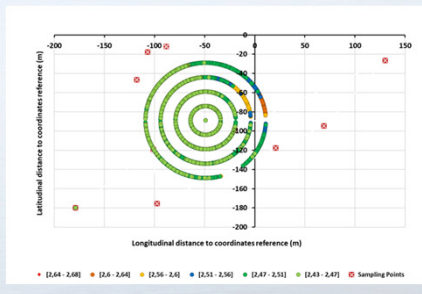









### Sampling Uncertainty Modelling

**Methodology:**

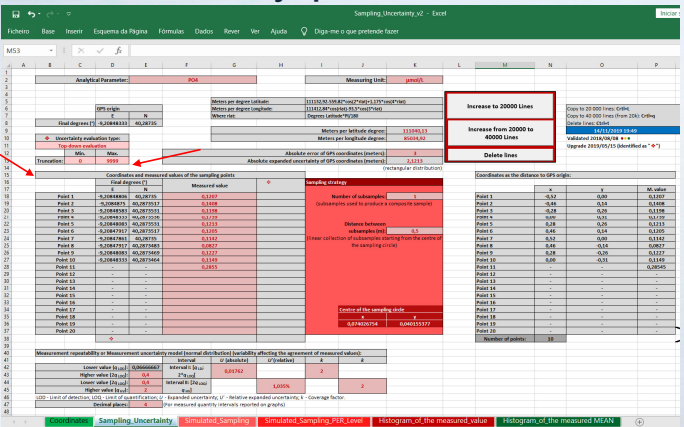
- 1) Collected samples at a defined depth and known GPS coordinates
- 2) Describe collected information as a 3D surface (x, x, z), where x and y are GPS coordinates and z samples concentrations
- 3) Randomization of the 3D surface given GPS coordinates and sample analysis precision
- 4) Random scanning of the randomised surfaces to model concentration variation (Single Sampling)
- 5) Simulate other types of sampling strategies










## Sampling Modelation

**The sampling modelling was based on Monte Carlo Simulations implemented in a user-friendly spreadsheet:**



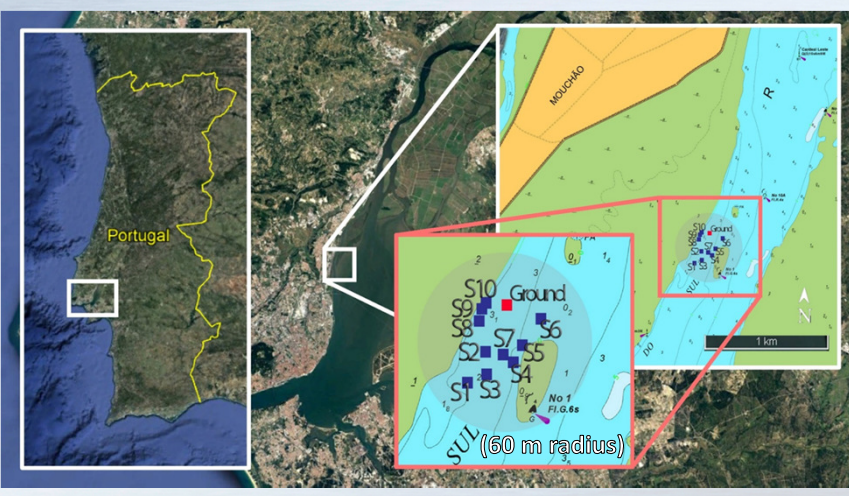
C. Borges, C. Palma, R. J. N. Bettencourt da Silva, *Optimization of river sampling: application to nutrients distribution in Tagus river estuary*, Anal. Chem. **91** (2019) 5698-5705.


## Sampling Procedure

### Estuarine Waters (Tagus River)

**Collection of 10 samples in a sampling circle of the river:**









### Sampling Procedure


#### Estuarine Waters (Tagus River)

Collection of 10 samples in a sampling circle of the river:



Niskin Bottle


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### Sampling Procedure




#### Ocean Waters (Portuguese Continental Platform)

Collection of 9 samples within a 10 x 10 nm grid;







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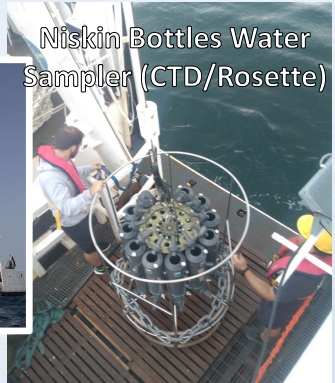




## Sampling Procedure

Ocean Waters (Portuguese Continental Platform)




Collection of 11 samples within a 1 m radius



Niskin Bottles Water Sampler (CTD/Rosette)

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## Results

The case of Phosphate

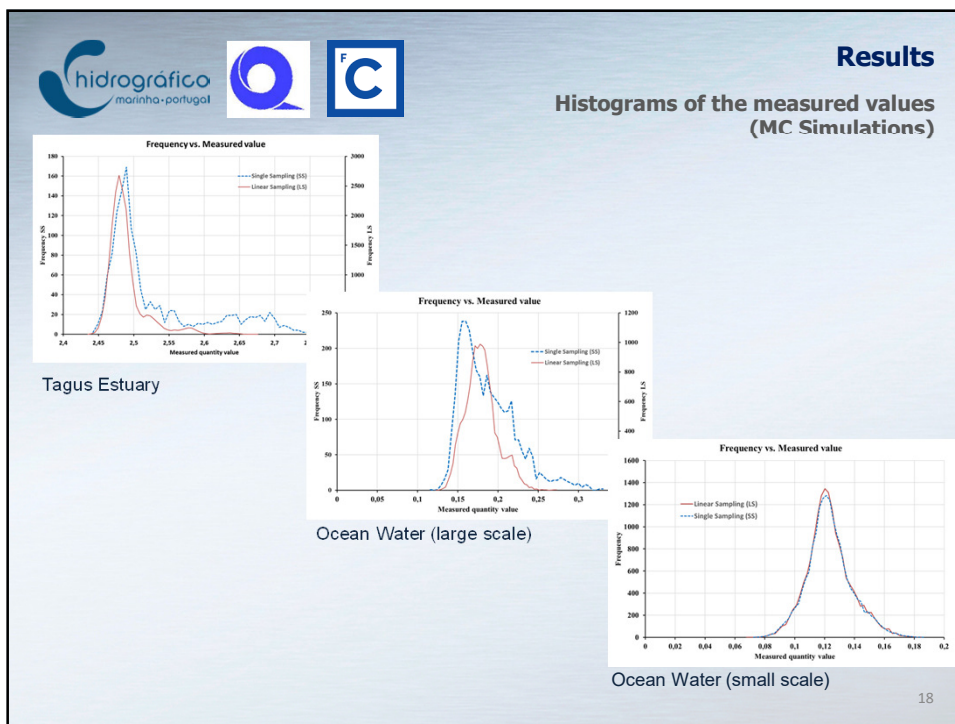
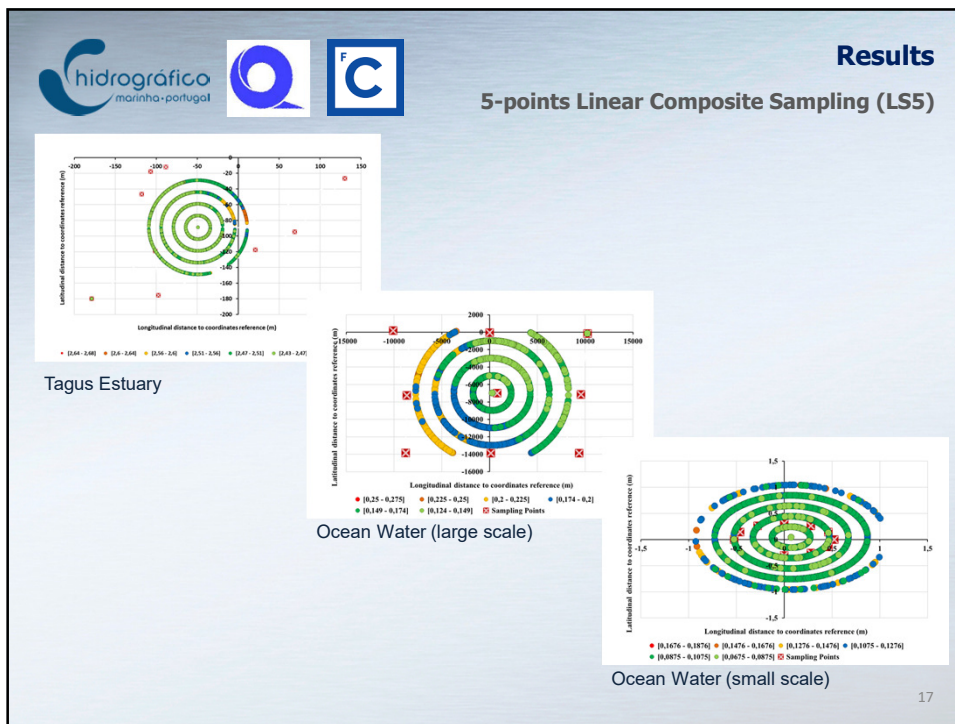
Simulated variability of the mass concentration and estimated sampling and combined expanded uncertainties using different sampling strategies.

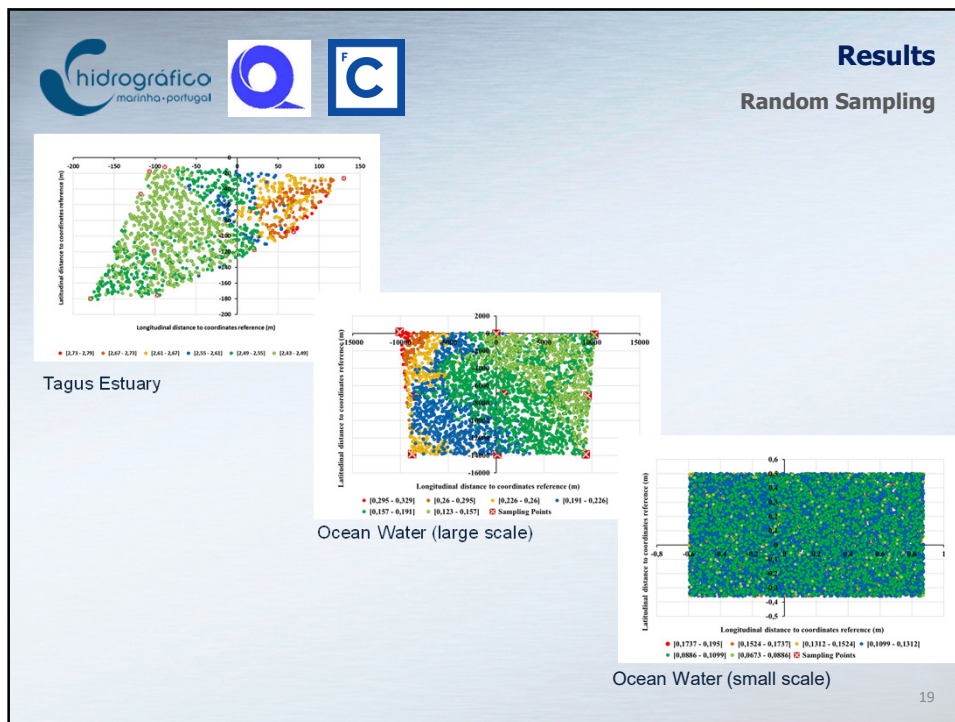
Sampling	Mean $\bar{s}$	$s'_r$ (%)	$s'_v$ (%) $\bar{s}$	$s'_i$ (%)	$u'_T$ (%)	$U'$ (%)
Tagus Estuary						
SS	3.08	0.52	2.26	2.35	3.21	9.21
RS(5)	-	0.52	1.01	2.35	3.21	8.28
LS(5; 15)	3.07	0.52	0.51	2.35	3.21	8.09
Ocean Water (large scale)						
SS	0.186	4.73	17.37	7.13	3.21	39.25
RS(5)	-	4.73	7.77	7.13	3.21	23.98
LS(5; 2000)	0.185	4.85	7.86	7.13	3.21	24.20
Ocean Water (small scale)						
SS	0.123	7.17	9.59	7.13	3.21	28.60
RS(5)	-	7.17	4.29	7.13	3.21	22.88
LS(5; 0.20)	0.123	7.15	4.47	7.13	3.21	23.00

$\bar{s}$  - Value obtained by the Monte Carlo Method; SS - Single sampling; RS(m) - Random composite sampling from m subsamples; LS(m; d) - Linear composite sampling from m subsamples positioned at d meters.

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**Conclusions**

The developed methodology for uncertainty modelling and optimisation was successfully applied to marine systems sampling;

**Major Features:**

- **Case 1 (Tagus Estuary):**
  - Duplicate or triplicate composite sampling (random or linear with 30 m distances) allow a useful reduction of the measurement uncertainty.
- **Case 2 (Ocean Waters, large scale):**
  - LS can not be applied to estimate sampling uncertainty due to system's heterogeneity
  - The use of the tool can be used to assess trends e detect abnormalities in large systems behaviour
- **Case 3 (Ocean Water, small scale):**
  - Duplicate or triplicate composite sampling (from distinct bottles) allow a useful reduction of the measurement uncertainty
- **Special Feature:**
  - Application of sampling uncertainty to establish different cargo capabilities scenarios

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