

**Sensitivity analysis to evaluate a new spatialized
process-oriented model of water and pesticide
transfers at the catchment scale**

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Introduction

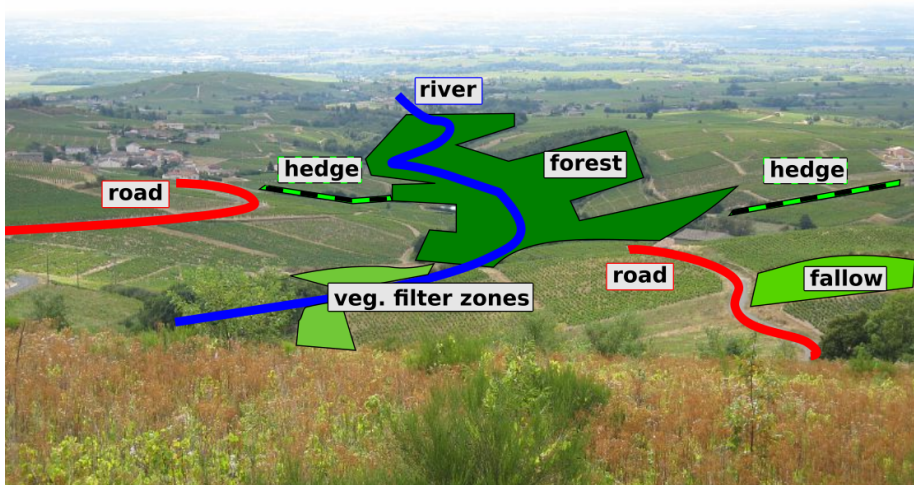
Context



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Introduction

Context

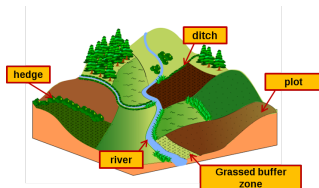


Introduction

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How to tackle pesticide transfers and fate on small agricultural catchments with modelling tools ?

- ✓ Integrating landscape elements diversity

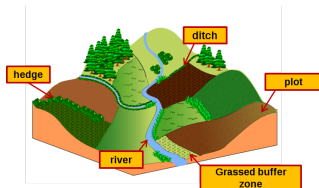


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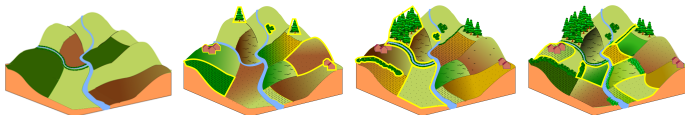
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- ✓ Exploring landscape management scenarios

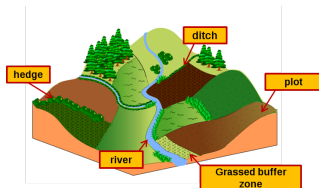


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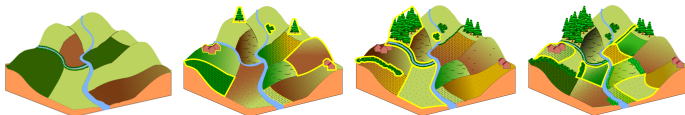
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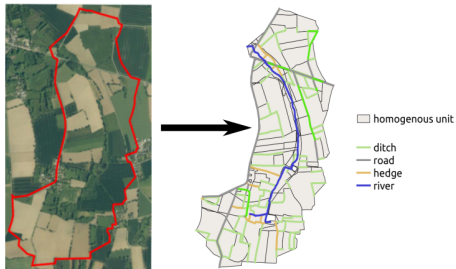


⇒ Development of the **PESHMELBA** model (Rouzies et al. 2019)

Introduction

The PESHMELBA model

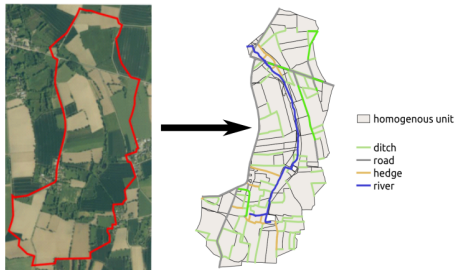
- ✓ Simulation of heterogenous landscapes composed of plots, vegetative filter zones, hedges, ditches and rivers



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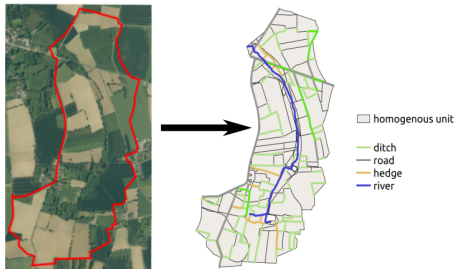
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- ✓ Water transfers on surface and subsurface



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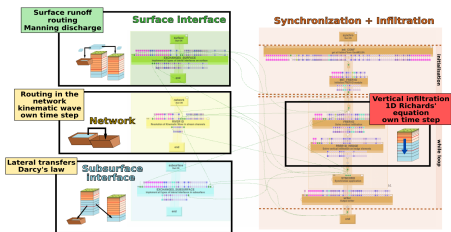
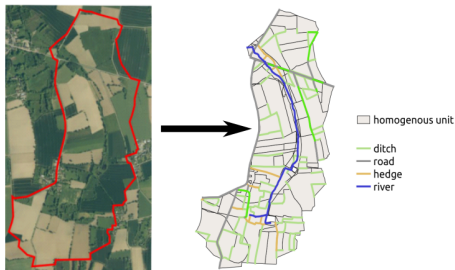
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- ✓ Water transfers on surface and subsurface
- ✓ Solute advection, adsorption and degradation



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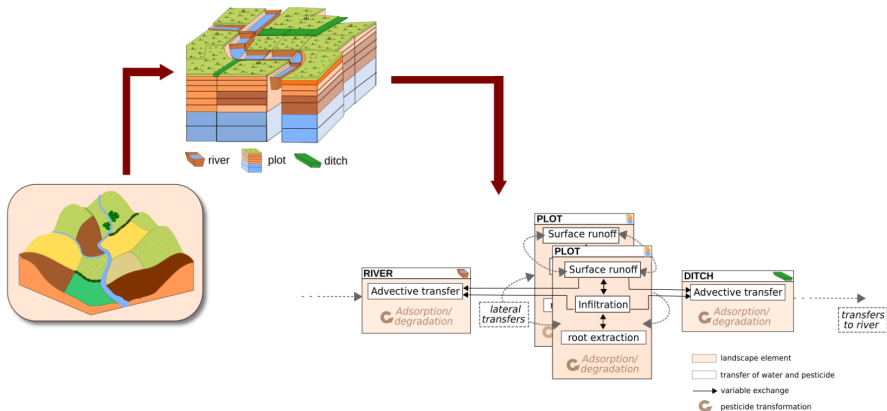
The PESHMELBA model

- ✓ Simulation of heterogenous landscapes composed of plots, vegetative filter zones, hedges, ditches and rivers
- ✓ Water transfers on surface and subsurface
- ✓ Solute advection, adsorption and degradation
- ✓ One module \equiv one process or ensemble of processes on a landscape element
- ✓ Coupling of modules within the OpenPALM coupler (Fouilloux and Piacentini 1999) turning the structure flexible



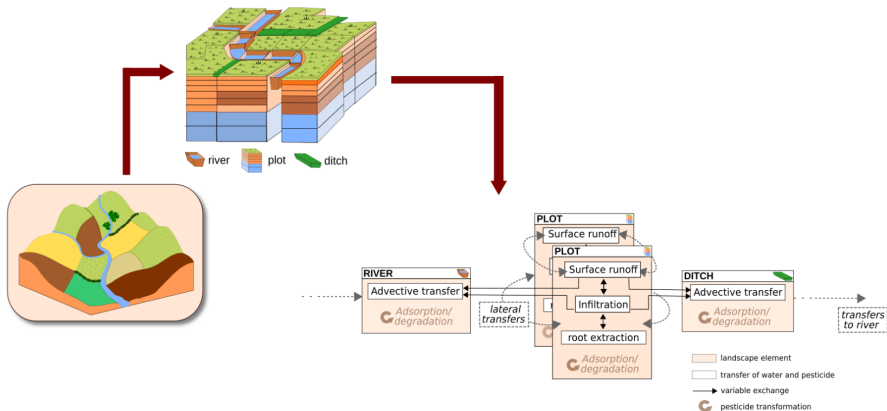
Introduction

The PESHMELBA model



Introduction

The PESHMELBA model



⇒ Complex structure may lead to additional difficulties to diagnose model behavior!

Introduction

Objectives



Necessary to quantify and reduce the uncertainty associated to the model output variables, particularly for decision-making.

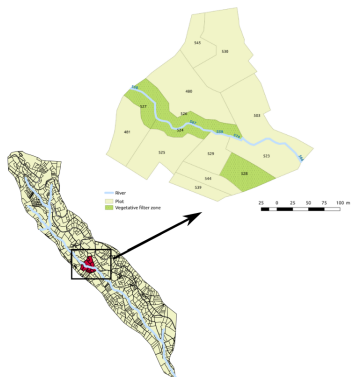
Global Sensitivity Analysis (GSA) is a necessary step.

- ? Which tools to address the question of GSA in a **modular, spatialized** model of **pesticide transfers**?
- ? How can GSA be a tool to assess physical processes representation in PESHMELBA model ?

Material and methods

Study case

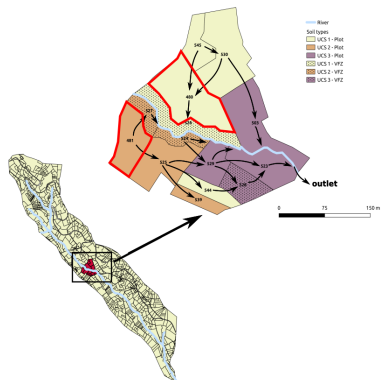
La Morcille catchment (France)



Material and methods

Study case

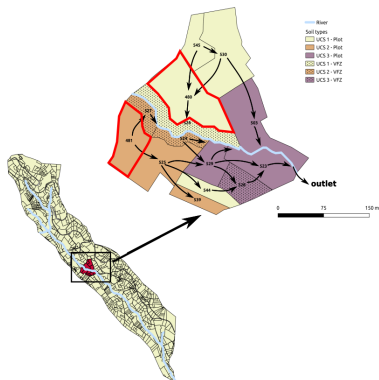
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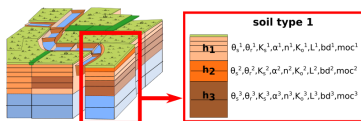
Material and methods

Study case

La Morcille catchment (France)



3 soil types : 15 horizons



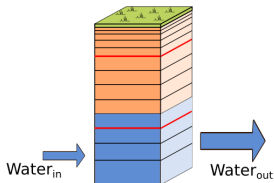
Soil	Plots/VFZ	River	Vegetation	Pesticide
thetas	hpond	hpond	manning	Kf _{oc}
thetar	adsorpthick	di	Zr	DT50
K _S		Ks	F10	
alpha		manning	LAI _{max}	
n				
Ko				
L				
bd				
moc				

⇒ **145 parameters to be sampled**
 but simulations computationally costly :
 limited to Latin Hypercube Sample of 4,000
 points

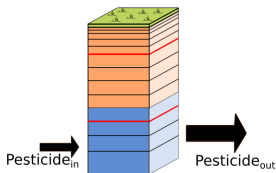
Material and methods

Target output

- ✓ Cumulated water lateral flow (saturated flow only)



- ✓ Cumulated pesticide lateral flow (advection/saturated flow only)



✓ **Variance-based Sobol method** (Sobol 1993)

Decomposition of the output variance in conditional variances.

⚠ Sobol method requires a very large sample that cannot be computed

⇒ Use of **Polynomial Chaos Expansion** (PCE) : surrogate method that provides Sobol indices

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✓ **HSIC dependence measure** (Da Veiga 2015)

Describes similarity between P_Y and $P_{Y|X}$ using a dependence measure d :

$$S_i^d = \mathbb{E}_{X_i}(d(P_Y, P_{Y|X_i}))$$

⇒ dependence measure d : Hilbert-Schmidt independence criterion (**HSIC**) : generalizes notion of **covariance** between two random variables

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Strategy :

- 1 Screening using HSIC dependence measure
- 2 Ranking using Sobol method
- 3 Comparison with ranking from HSIC dependence measure

Spatial aspects

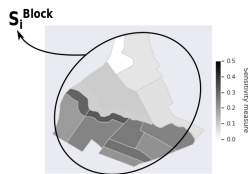
Block vs site indices (Saint-Geours 2012)



2 distinct ways to scrutinized spatialized output variables sensitivity :



1. **Site sensitivity indices:** as many analysis as spatial locations



2. **Block sensitivity indices:** Single analysis with respect to the whole spatial domain, aggregated indices

Results

Screening

Reminder : 145 parameters initially sampled

Soil	Plots/VFZ	River	Vegetation	Pesticide
thetas thetar K_s alpha n Ko L bd moc	hpond adsorp thick	hpond di Ks mannings	mannings Zr F10 LAI _{max}	K _{foc} DT50

Results

Screening

Reminder : 145 parameters initially sampled

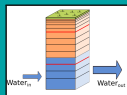
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After screening (statistical hypothesis tests based on HSIC measure (De Lozzo and Marrel 2014)):

- ✓ Cumulated Water Lateral Flow : **84 remaining parameters**
- ✓ Cumulated Pesticide Lateral Flow : **80 remaining parameters**

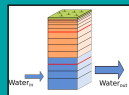
Results

Sobol - Water lateral flow



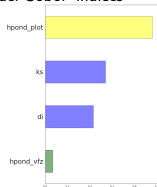
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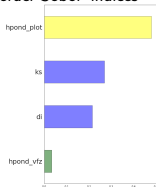


1. Block sensitivity

1st order Sobol' indices

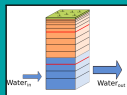


Total order Sobol' indices



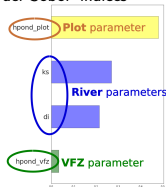
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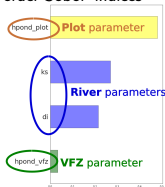


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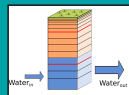
Total order Sobol' indices



High influence of **river parameters** and **ponding height** on plots and VFZs but...no interactions captured by PCE !?!

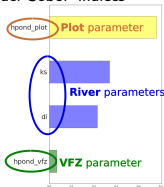
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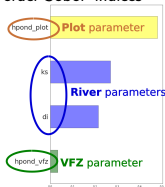


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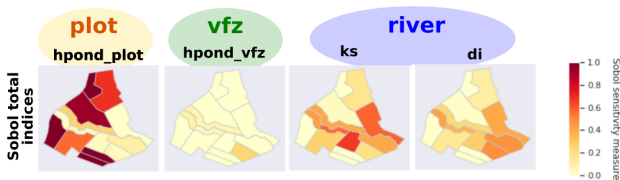


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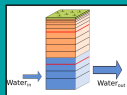
2. Site sensitivity



⇒ Spatial heterogeneities in influential parameters

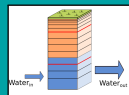
Results

HSIC - Water lateral flow



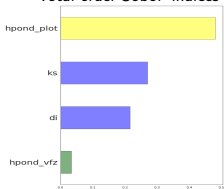
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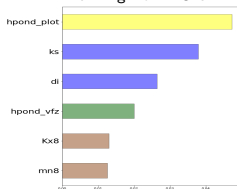


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Total order Sobol' indices

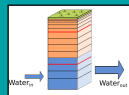


Ranking from HSIC

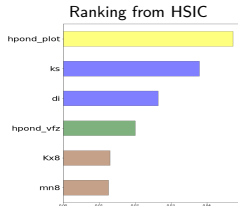
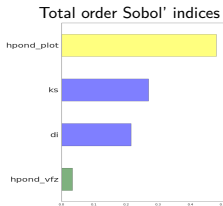


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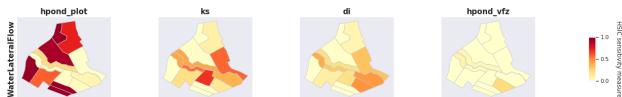
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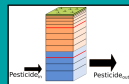
2. Site sensitivity



⇒ Good match between HSIC and Sobol ranking

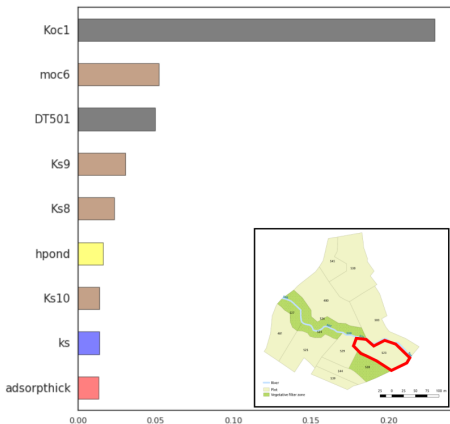
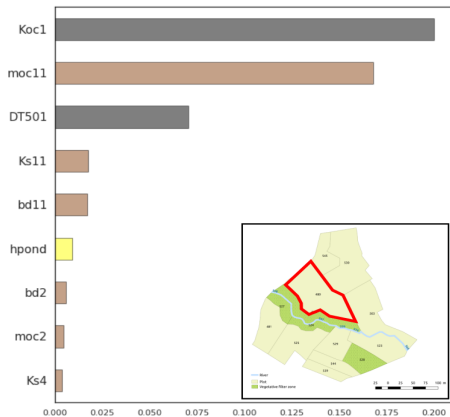
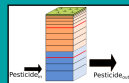
Results

HSIC - Pesticide lateral flow



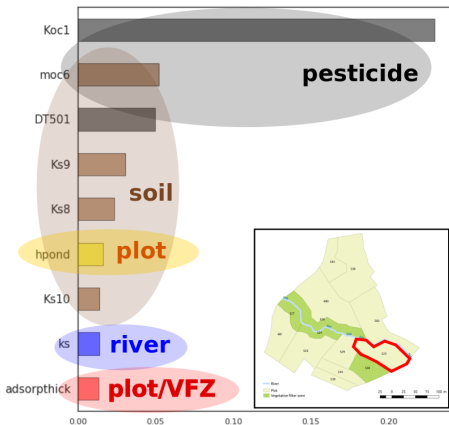
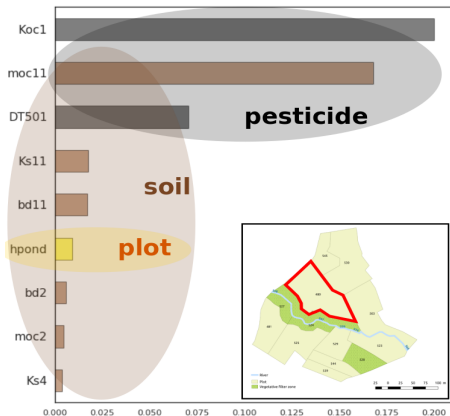
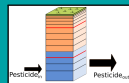
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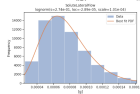
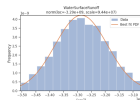
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Conclusion

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- ✓ Sobol' indices hard to estimate on PESHMELBA model
- ✓ HSIC measure as a consistent alternative for sensitivity measure
- ✓ A few parameters are identified as influential but spatially heterogeneous
- ✓ LHS also informs about output variables distribution : mainly gaussian or lognormal
⇒ Valuable information for next step : uncertainty reduction using **data assimilation**









Thanks for your attention !

Questions ?

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References

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