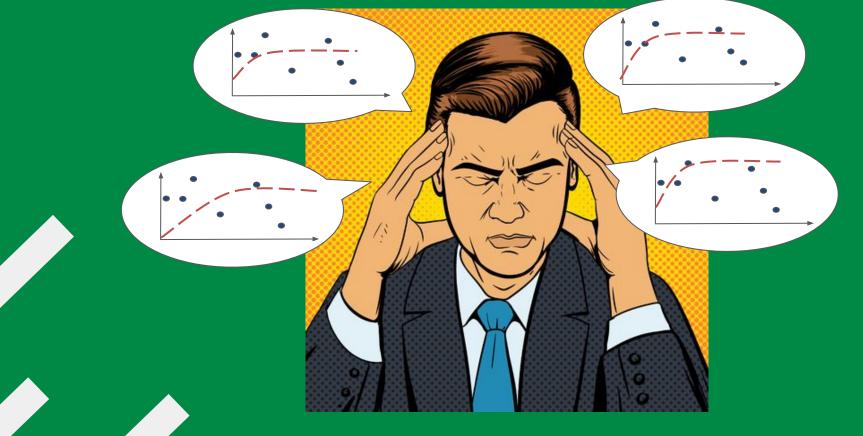
# Dealing with unfittable variograms

**Application of the GeoLime Sensitivity Analysis Module** 

## In books and class room variograms are easy to fit ...



## ... While things get more complex in real life



#### One can fit the same data with different variogram models depending on:



His

#### What do we usually do?



Use one acceptable variogram model and admit it is the best

\* Which we all have done

Unreasonably Tweak the experimental variograms, hide some pairs, and use a 10-structure model in a Gaussian space\*

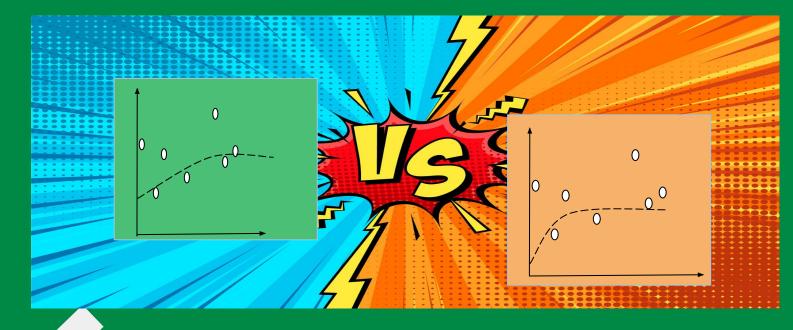
(e)

(1)

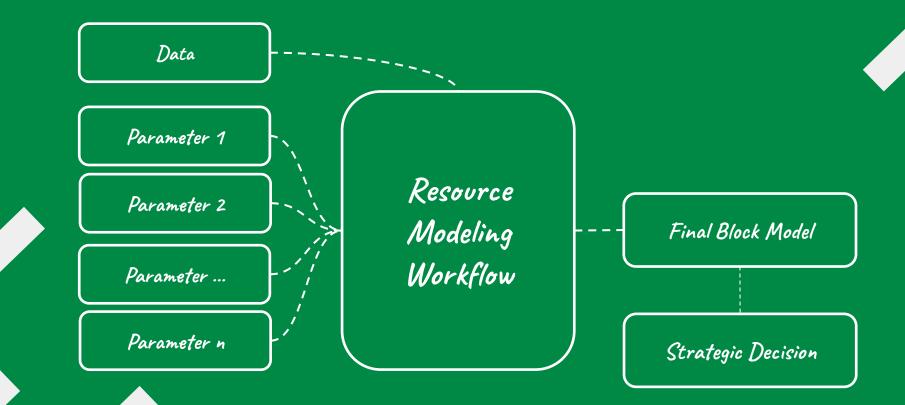
### Why is it a problem ?

Block model will be used for strategic decision based on the economic valuation of the deposit

#### Two differents, but similar, variogram models might leads to two economical decisions



#### **Deterministic modeling**



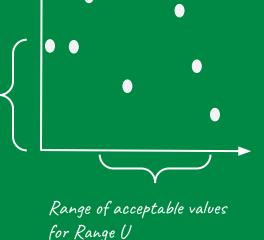
### The variogram dilemma

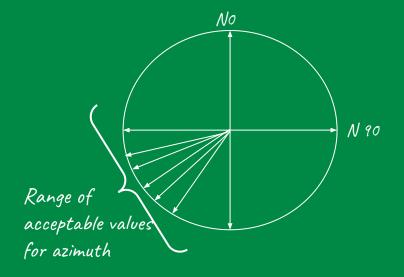
- If the variogram is uncertain, is the block model acceptable?
- Should one use the model with the highest spatial variability?
- So lower mining selectivity, and usually smallest
  - economical value?
  - Who knows what would happens if we used another nugget effect?

### The variogram dilemma

One cannot always be sure of the model to use







# How to solve the variogram dilemma ?

# First, identify the uncertain parameters

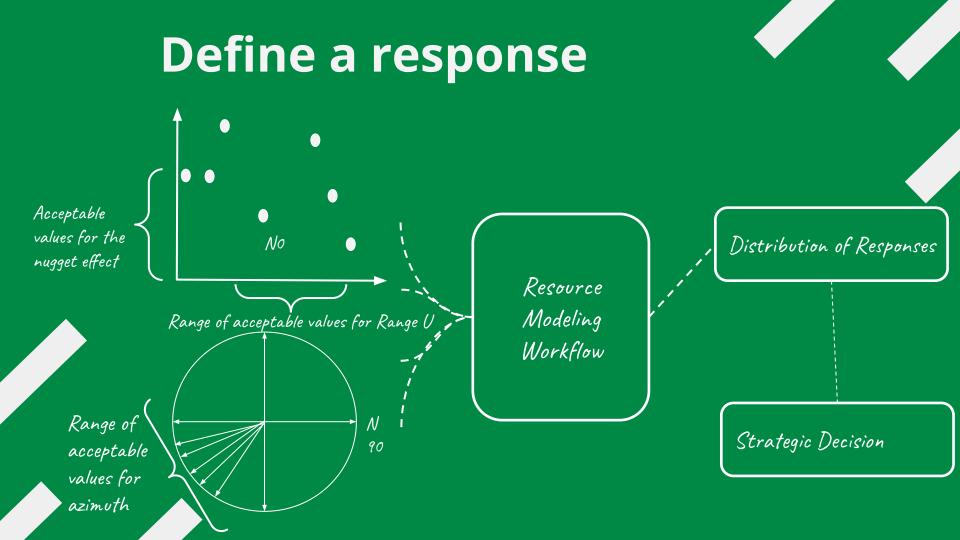


#### **Define a response**

Define a metric to evaluate the difference between the models: Ex: Total Tonne of ore at a specific cutoff

And understand the impact of the parameters on the response:

- How much each parameter impacts on the response?
- What would be the distribution of the response if we could test all possible variogram models?



#### **Distribution of Responses**

P90

Frequency P50 P10

Getting the distribution of the Ore Tonnage given the uncertainty over the variogram model, allow to analyze the risk on the strategic decision

Response (Ore Tonnage)

## The not-so-bad, but naive, solutions





With 6 parameters and 20 possibles value for each parameters there would be 64 millions models to test ! If the estimation procedure (kriging, mik, etc.) takes 20 min it would be 2500 years of computation Testing one variable at a time (OVAT) means the co-effect of parameters will be missed



### Teasing on GeoLime Solution

#### A better Approach: Sensitivity Analysis

Design a test plan to estimate the distribution of the response

Designing a test plan to reduce the 64 millions of model to 100 model s. As we want to get a distribution, we cannot just select a 100 models and apply the estimation workflow

#### A better Approach: Sensitivity Analysis

Model the surface response F(p1,p2,p3,...) = Total\_ore

> GeoLime library get has a dedicated function to test sensitivity on the ore volume from the variogram parameters

### Specialists behind GeoLime Sensitivity Analysis

#### Sebastien Strebelle, PhD:

- + World-renowned geostatistician
- + Former leader geoscientist R&D at Chevron for 20 years

#### **Claude Cavelius:**

- + Former R&D engineer at Chevron for 10 years
- + Former Lead Developer at Belmont Technology
- + Currently CTO at DeepLime

#### Christelle Lusso, PhD:

- + R&D in academia for 6 years
- + Former data engineer at Capgemini
- + Currently R&D in AI and mathematical R&D at DeepLime

Contact us to get a demo of the sensibility analysis with GeoLime

Contact@deeplime.io