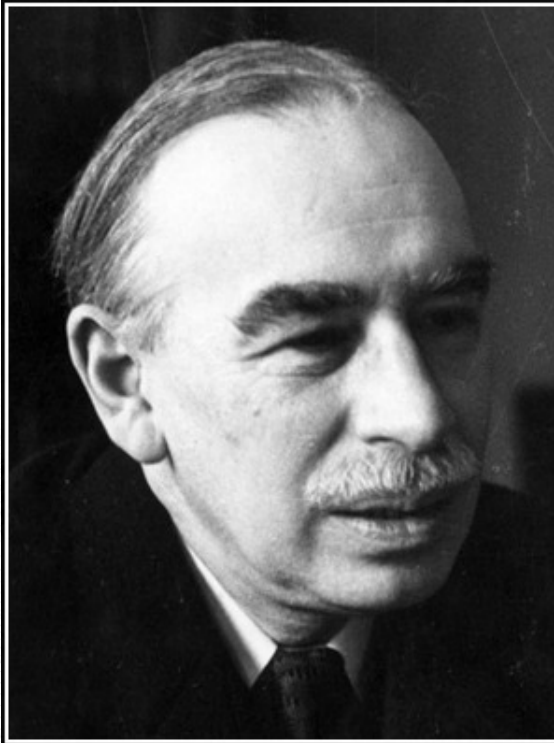


鲁棒优化和概率预测量化地质模型不确定性



Quantifying Geologic Uncertainty through Brownfield Optimization and Probabilistic Forecasting



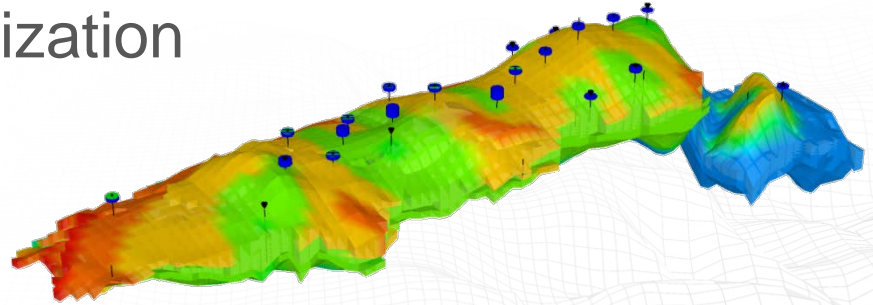
I'd rather be vaguely right than
precisely wrong.

— *John Maynard Keynes* —

AZ QUOTES

Agenda

- New Approach to History Matching
- Probabilistic Forecasting
- Introduction to Robust Optimization



Keynes' Challenge

**Would I Rather Have One Precise History
Match Or A Range Of
Plausible Solutions?**

Bayesian Formulation

$$P(\text{model}|\text{data}) = P(\text{model}) \frac{P(\text{data}|\text{model})}{P(\text{data})}$$

Posterior Probability

Prior Probability

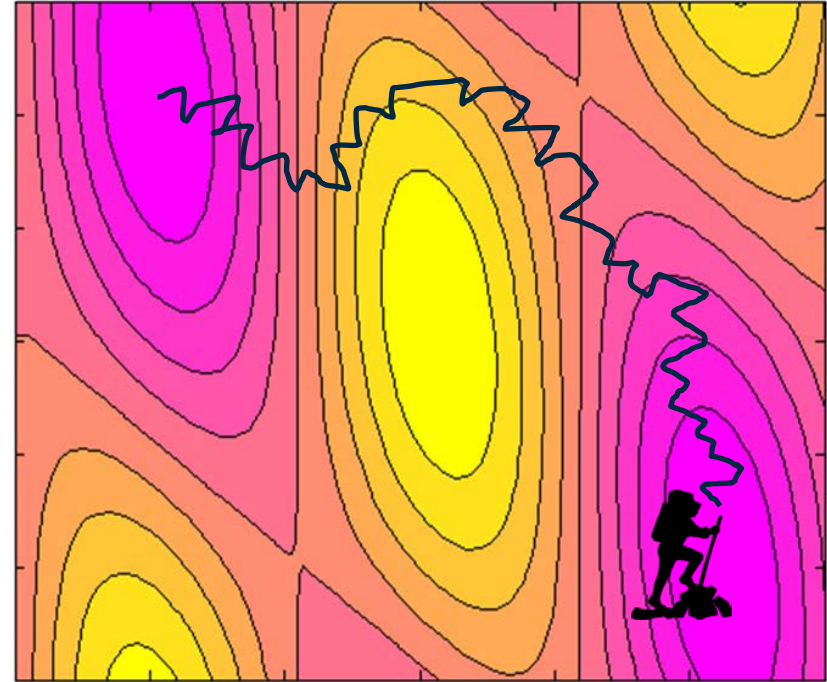
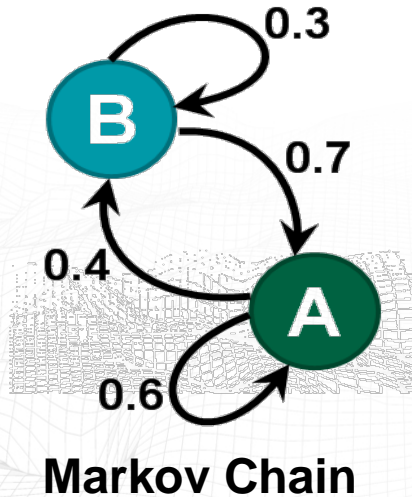
Conditional Probability

Thomas Bayes

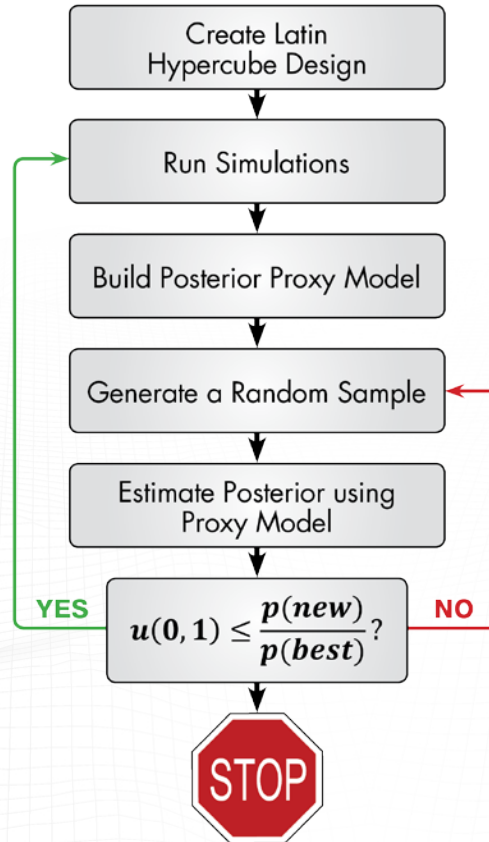


1701-1761

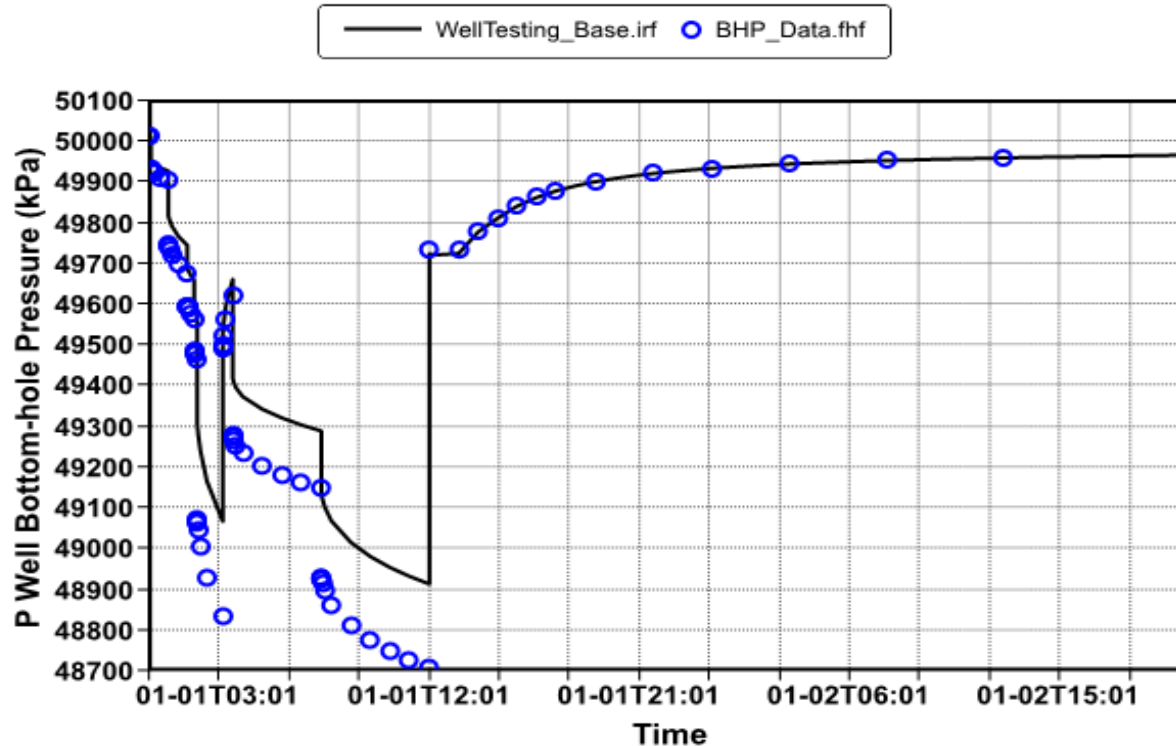
Metropolis-Hastings Markov Chain Monte Carlo (MCMC)



Proxy-based Acceptance-Rejection (CMG PAR)



Two Parameter Proof of Concept

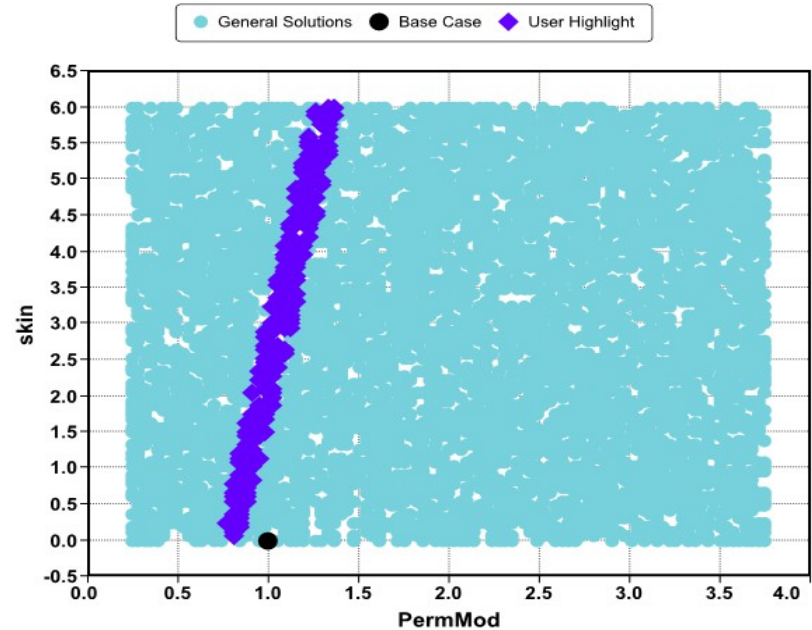
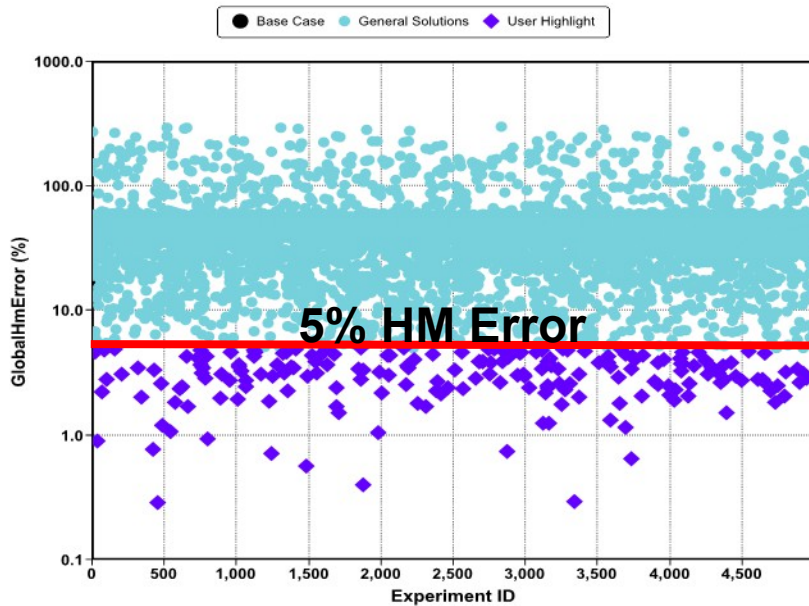


HM Parameters:

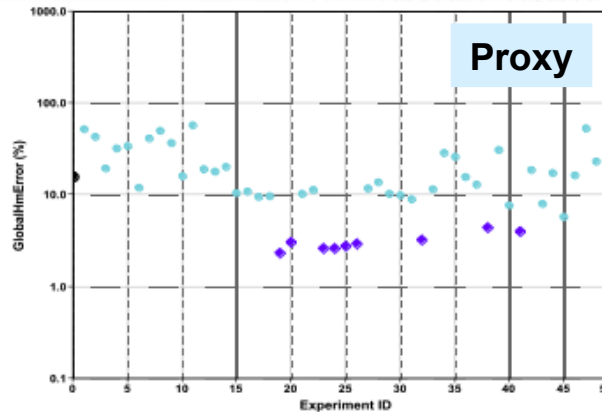
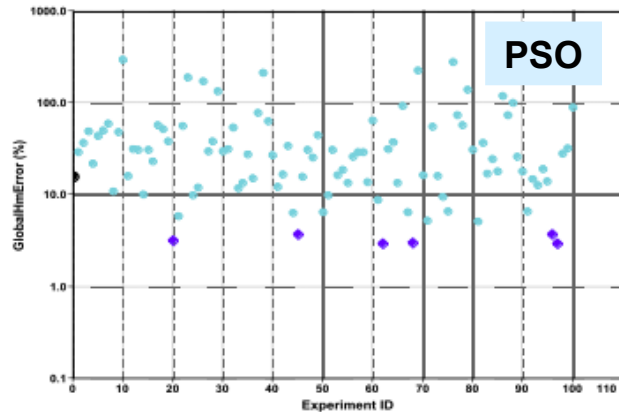
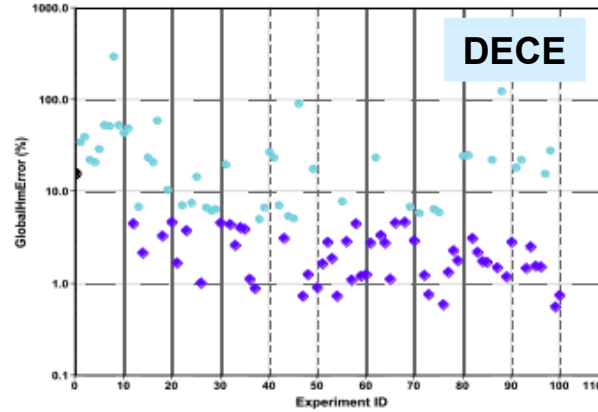
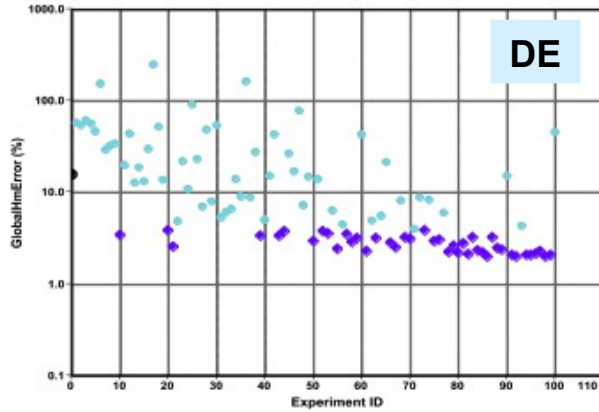
- Permeability
- Skin

$$PI = \frac{2\pi khff_h}{\ln \frac{r_e}{r_w} + S}$$

Reference Result – 5000 Brute Force Search

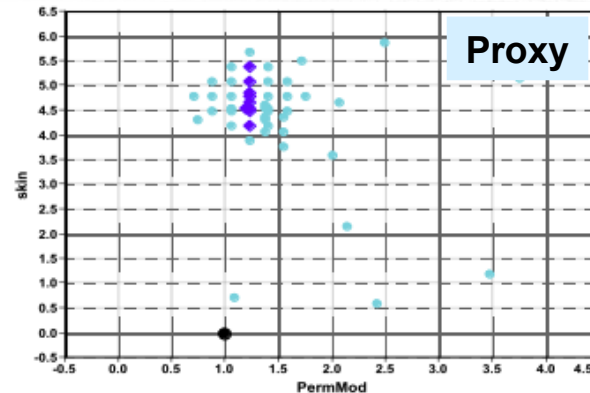
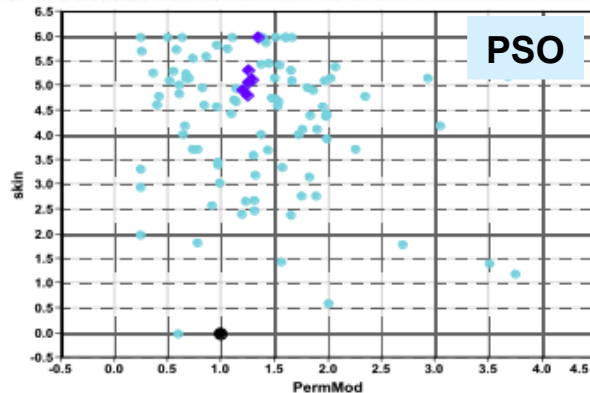
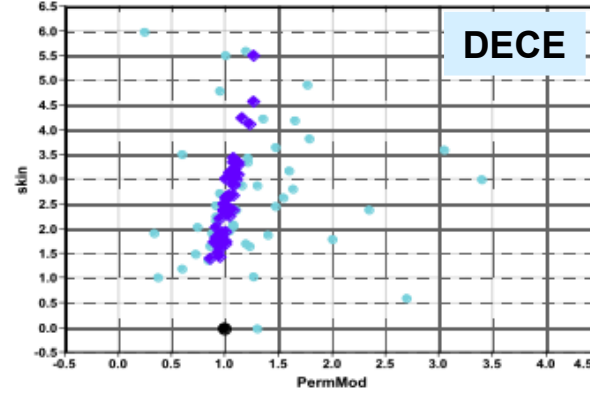
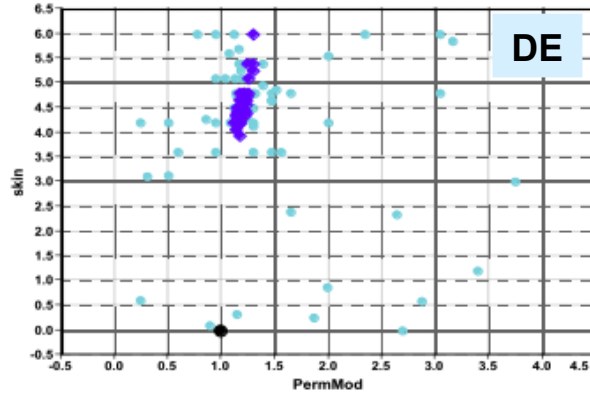


Results from Current Engines



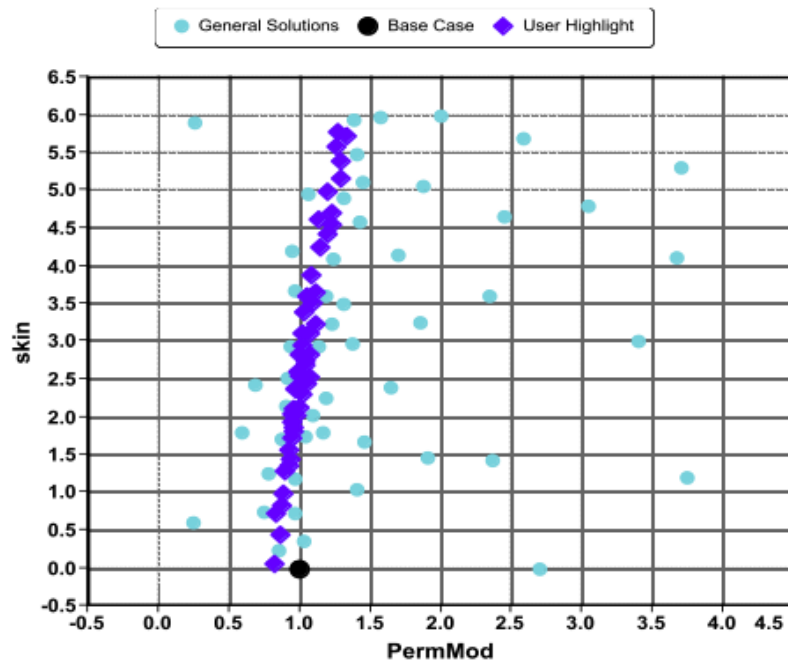
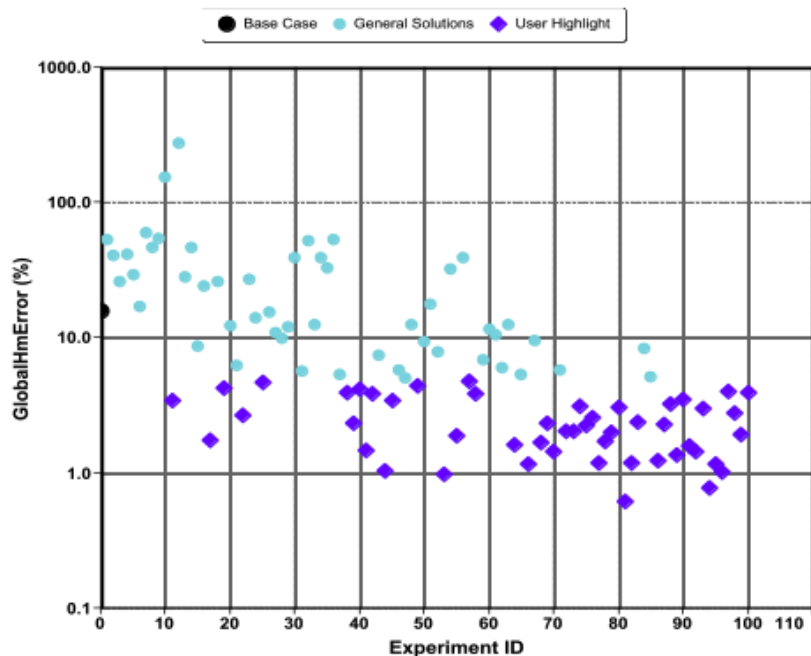
Each engine runs 100 jobs

Results from Current Engines

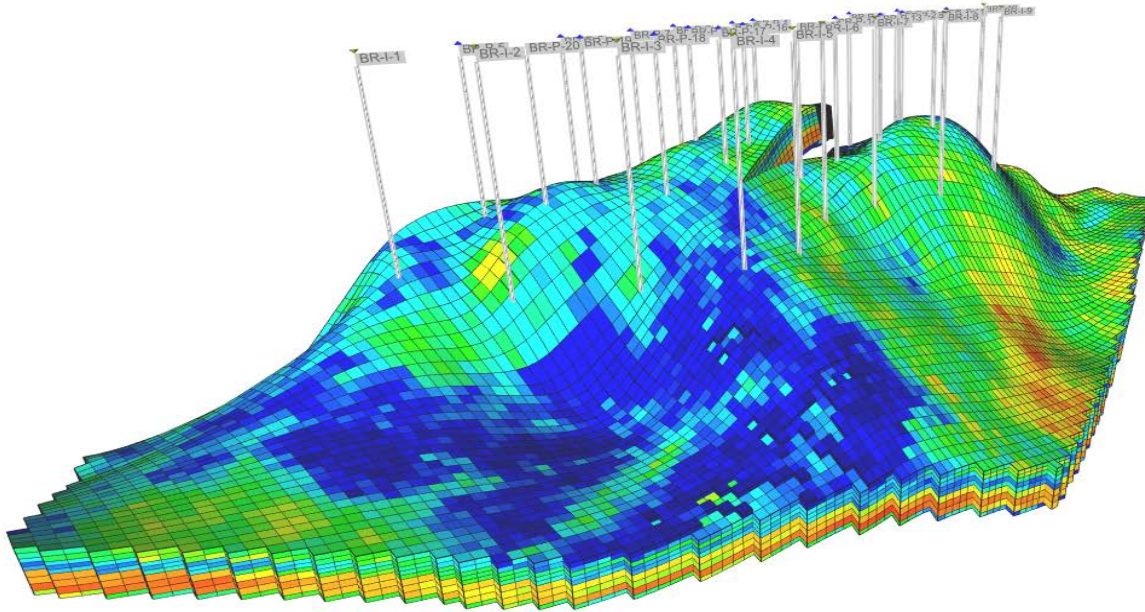


Each engine runs 100 jobs

Results from CMG PAR



Brugge HM Case Study



Courtesy of TNO

HM Workflow



**Rank and Select
Representative
Geologic
Realizations**



**Apply Additional
Uncertain
Parameters**



**Select History
Match
Parameters**

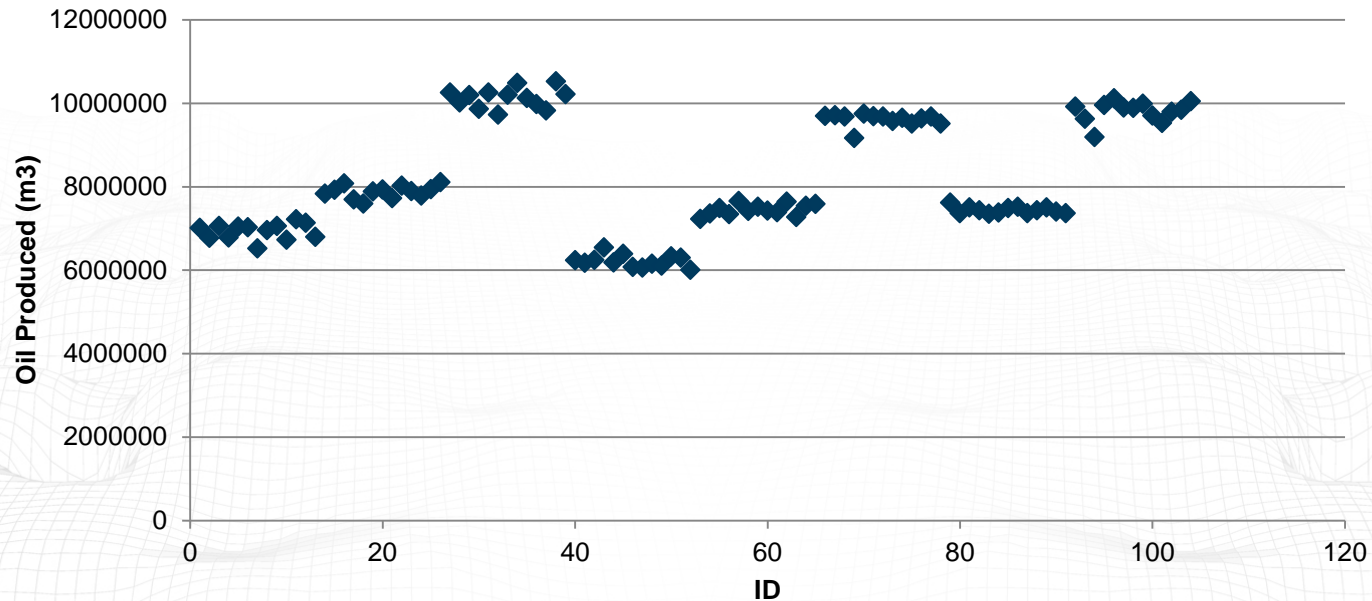


**Run CMG PAR
Engine**

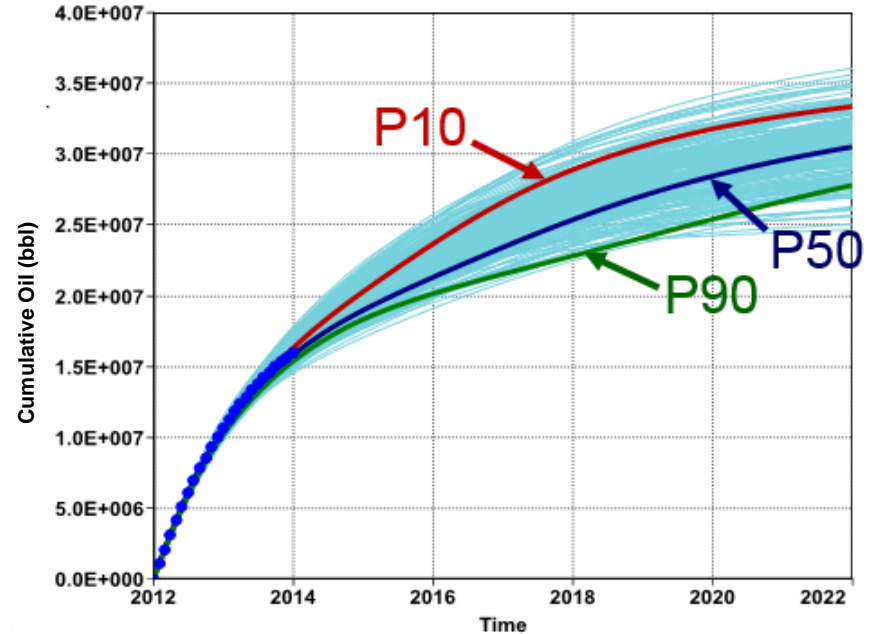
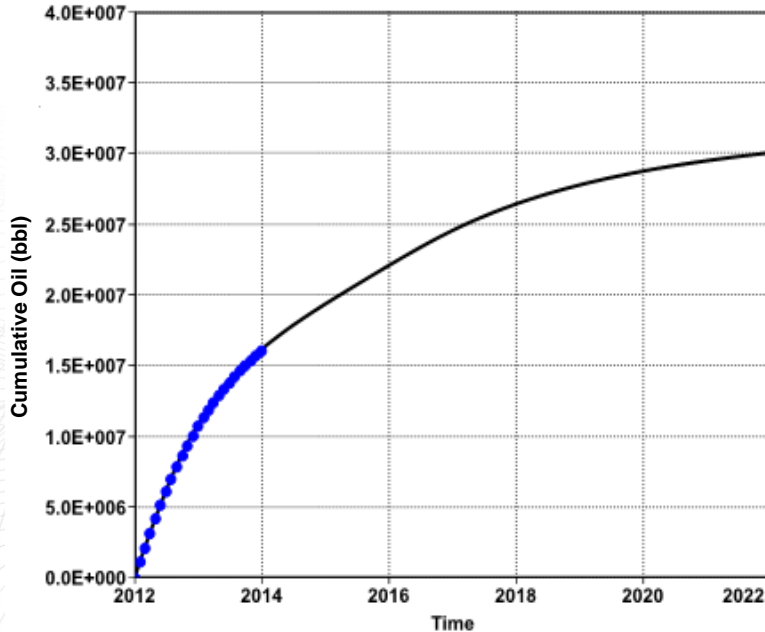
Rank Geologic Realizations



Cumulative Oil versus ID

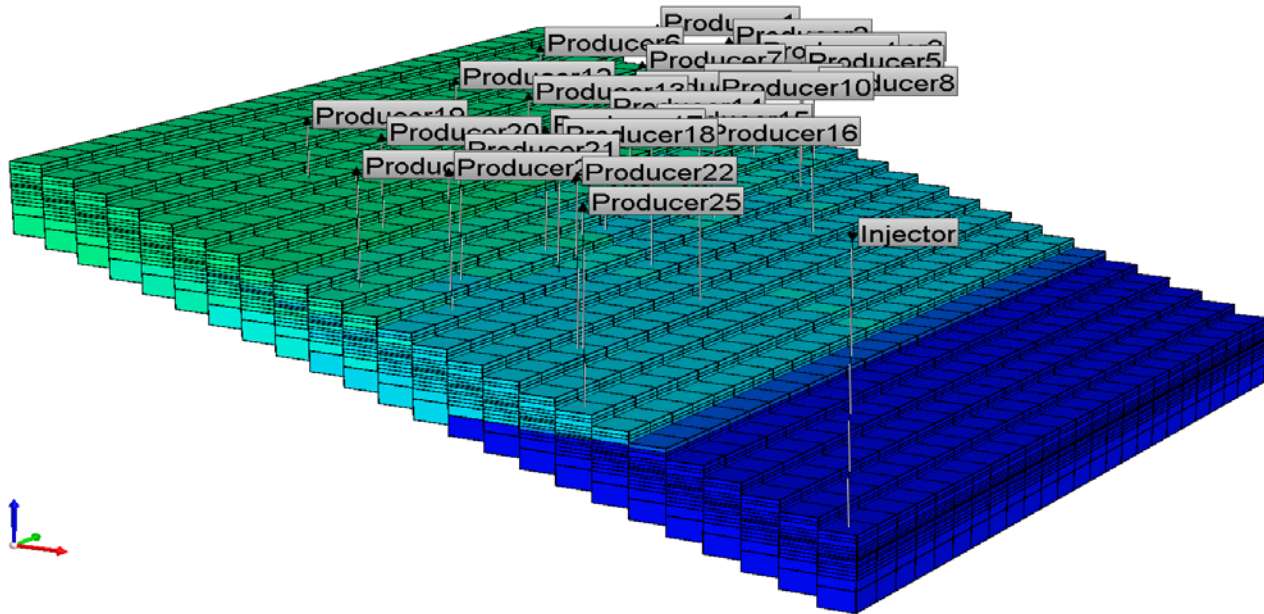


Probabilistic Forecasts



Probabilistic Forecast Proof of Concept

- 9th SPE comparative solution project
- 24x25x15 grid
- 1 water injection well, 12 producers



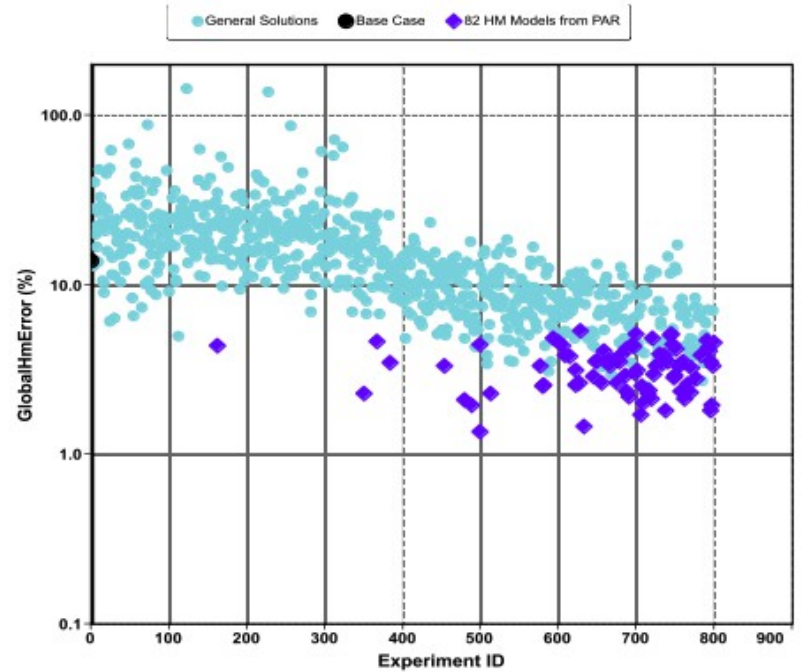
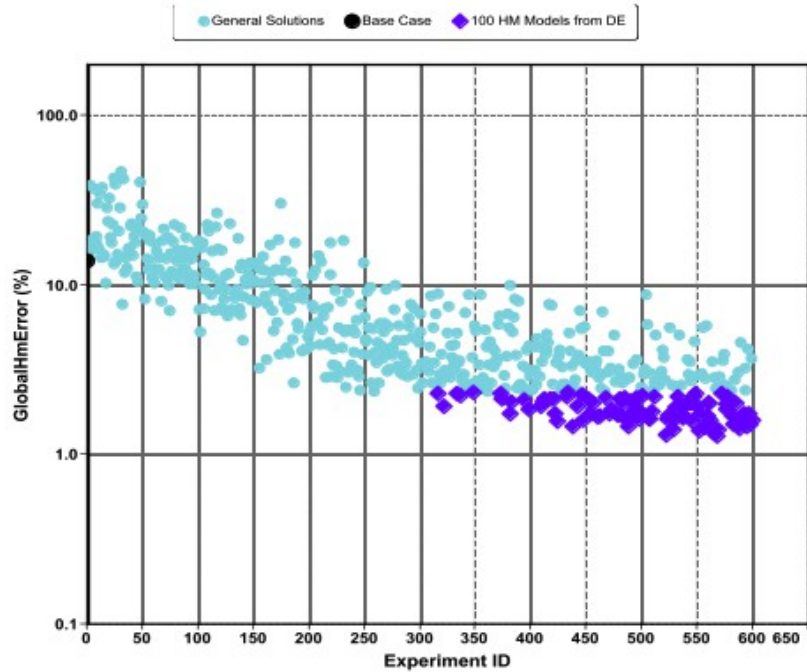
History Matching Parameters

- Permeability multipliers
 - 15 multipliers, one for each layer
- KV to KH ratio
 - One for entire model

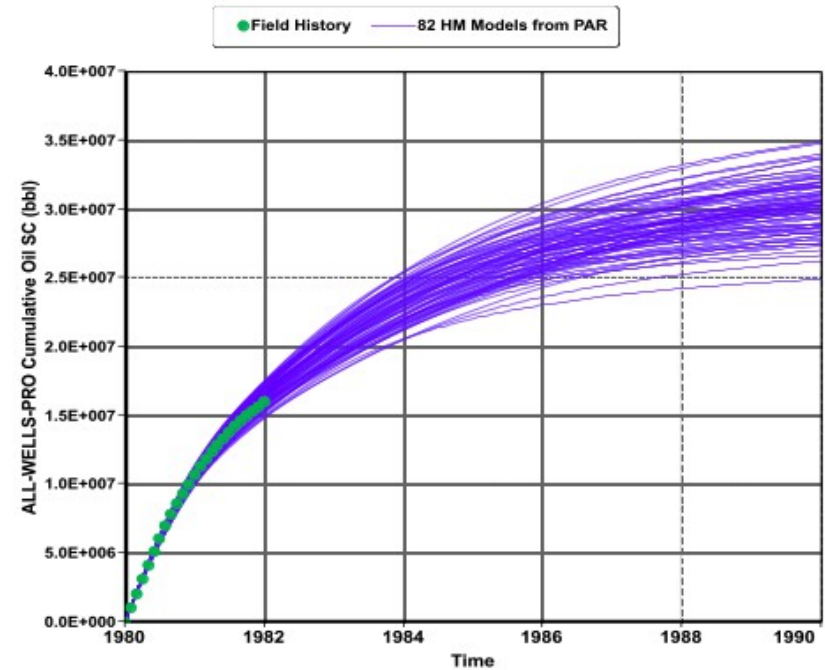
Summary of HM parameters

SWT and capillary pressure (end points and exponents)	10
SLT (end points and exponents)	6
Permeability multipliers	15
KV to KH ratio	1
Total	32

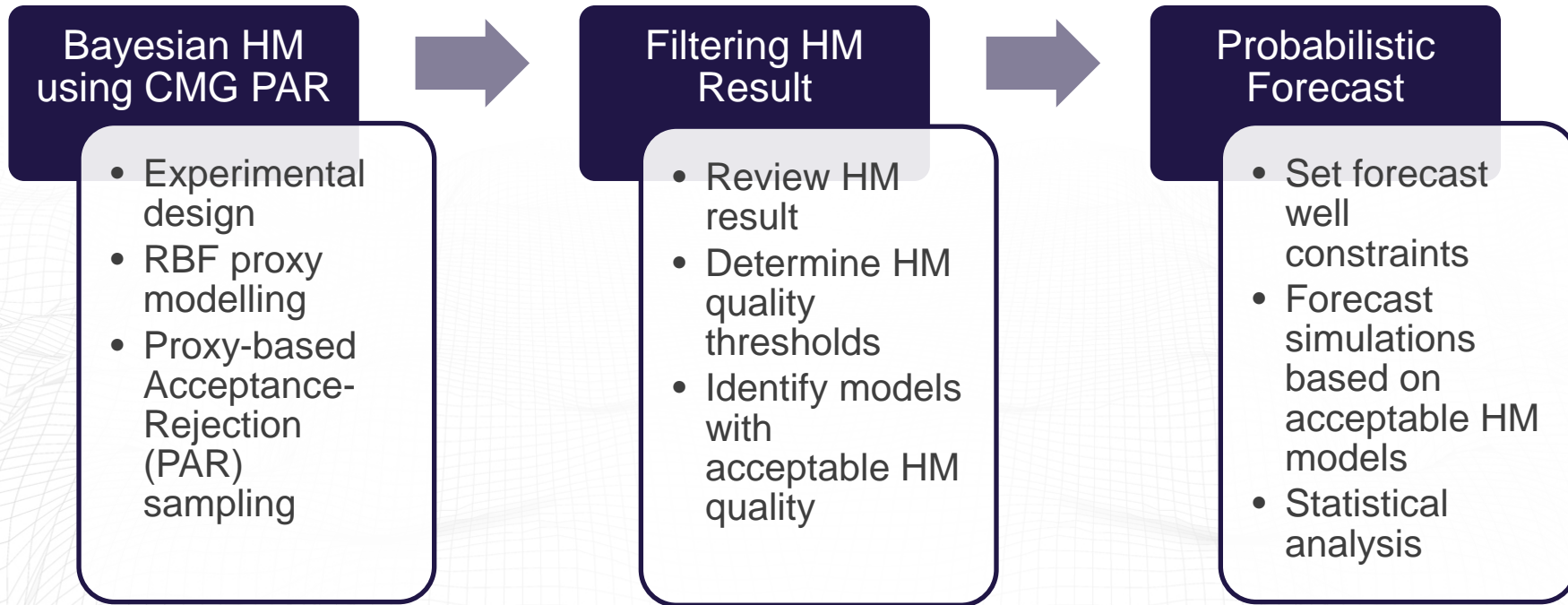
History Matching



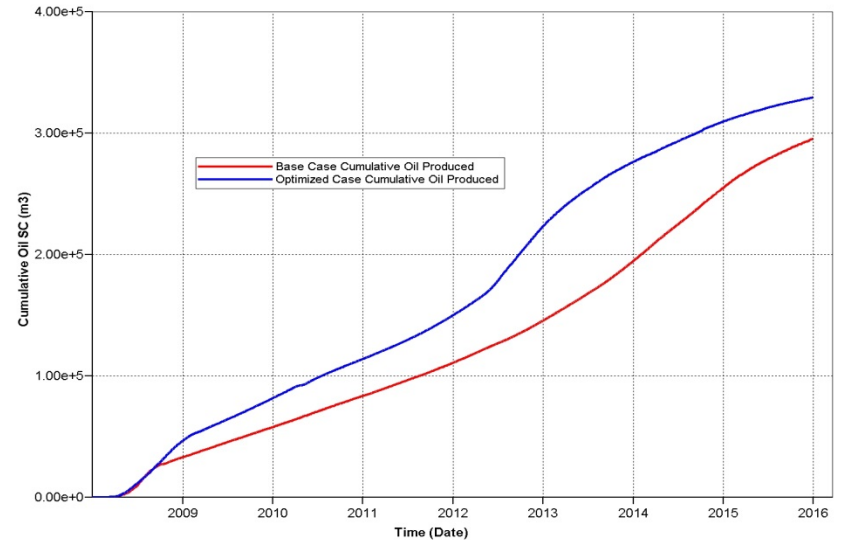
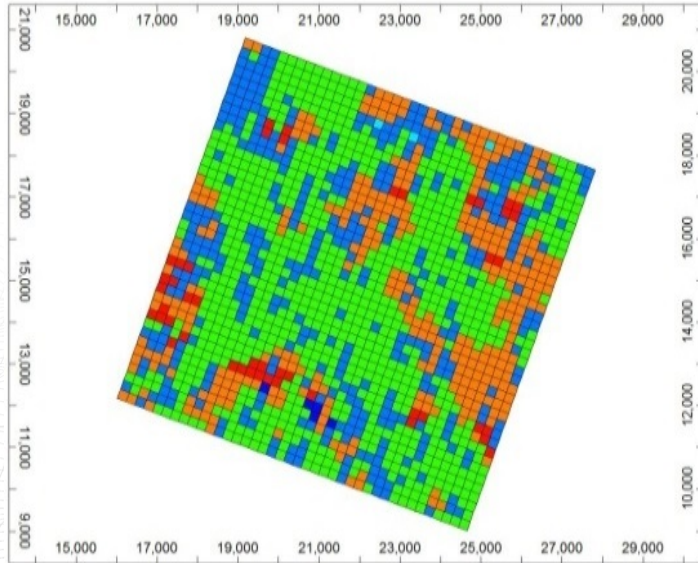
Probabilistic Forecast



Probabilistic Forecast Workflow using CMG PAR Sampling Method

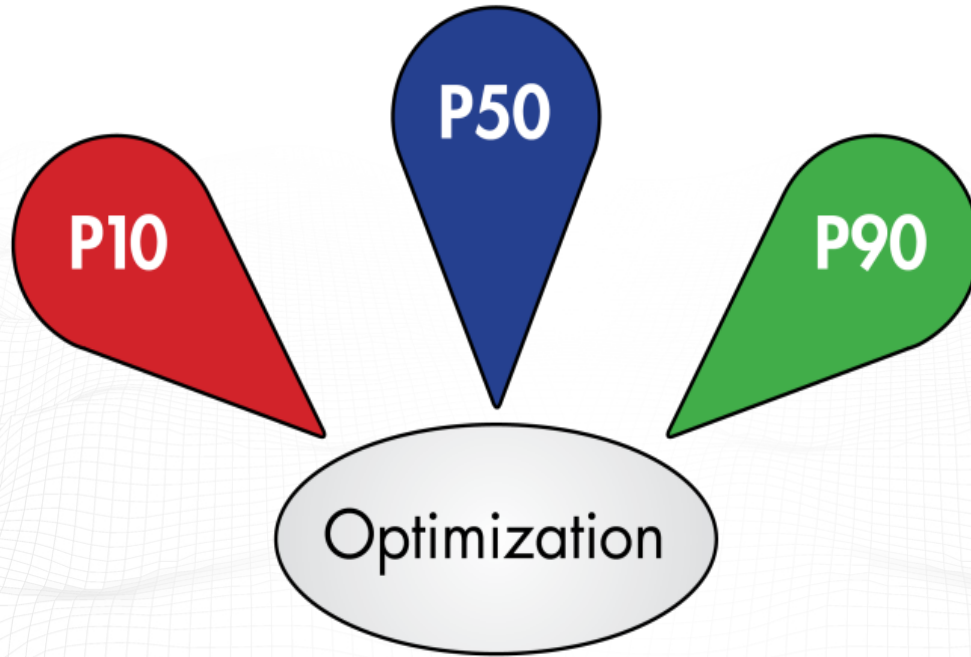


Traditional Nominal Optimization



Are we “precisely wrong?”

What is Robust Optimization?



Nominal vs. Robust Optimization



Nominal Optimization

Pros

- Fast: one simulation per scenario
- Likely to improve results

Cons

- No guarantee of success
- May be suboptimal in reality

Robust Optimization

Pros

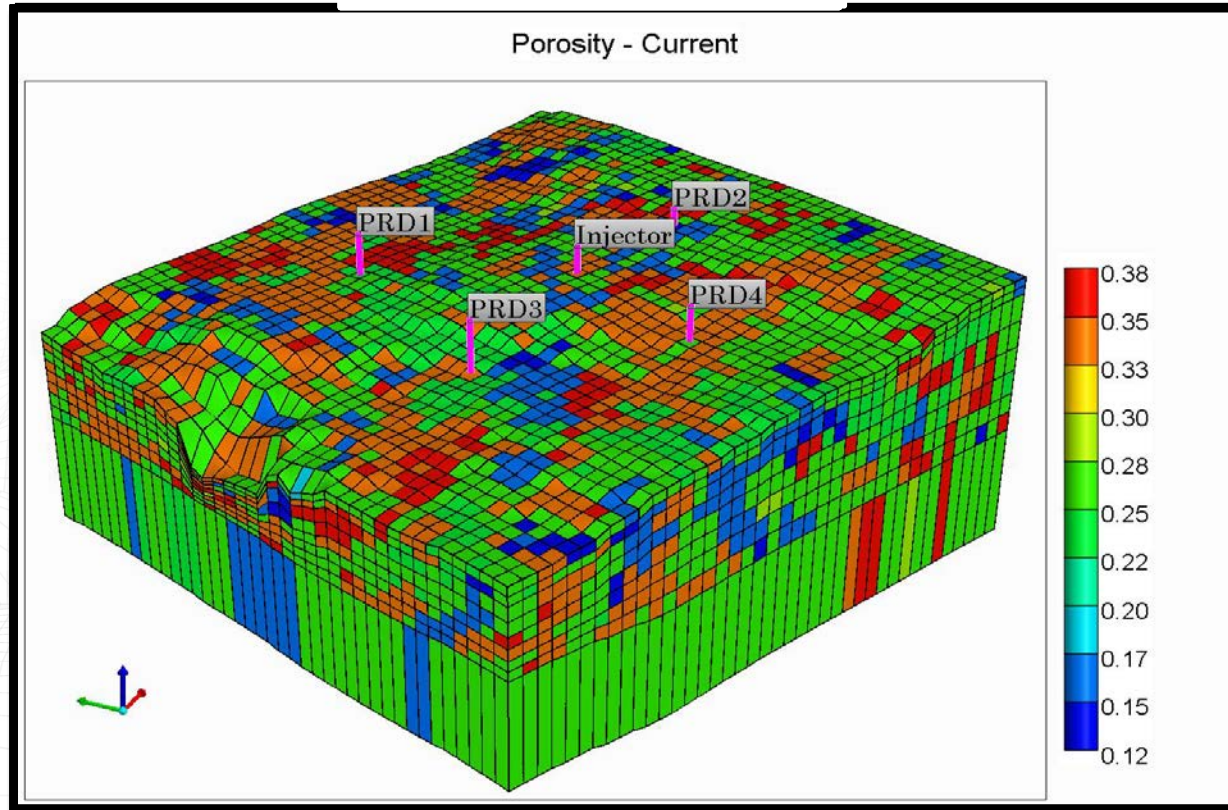
- Higher probability of success
- Lower risk

Cons

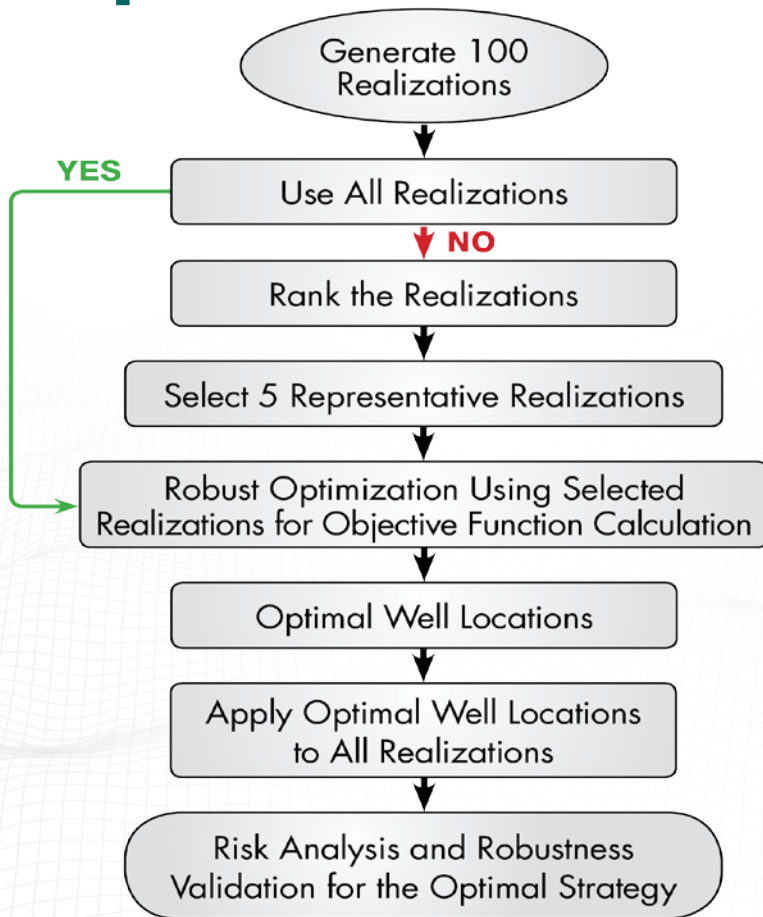
- Additional computation

Results: Optimal Wells Locations

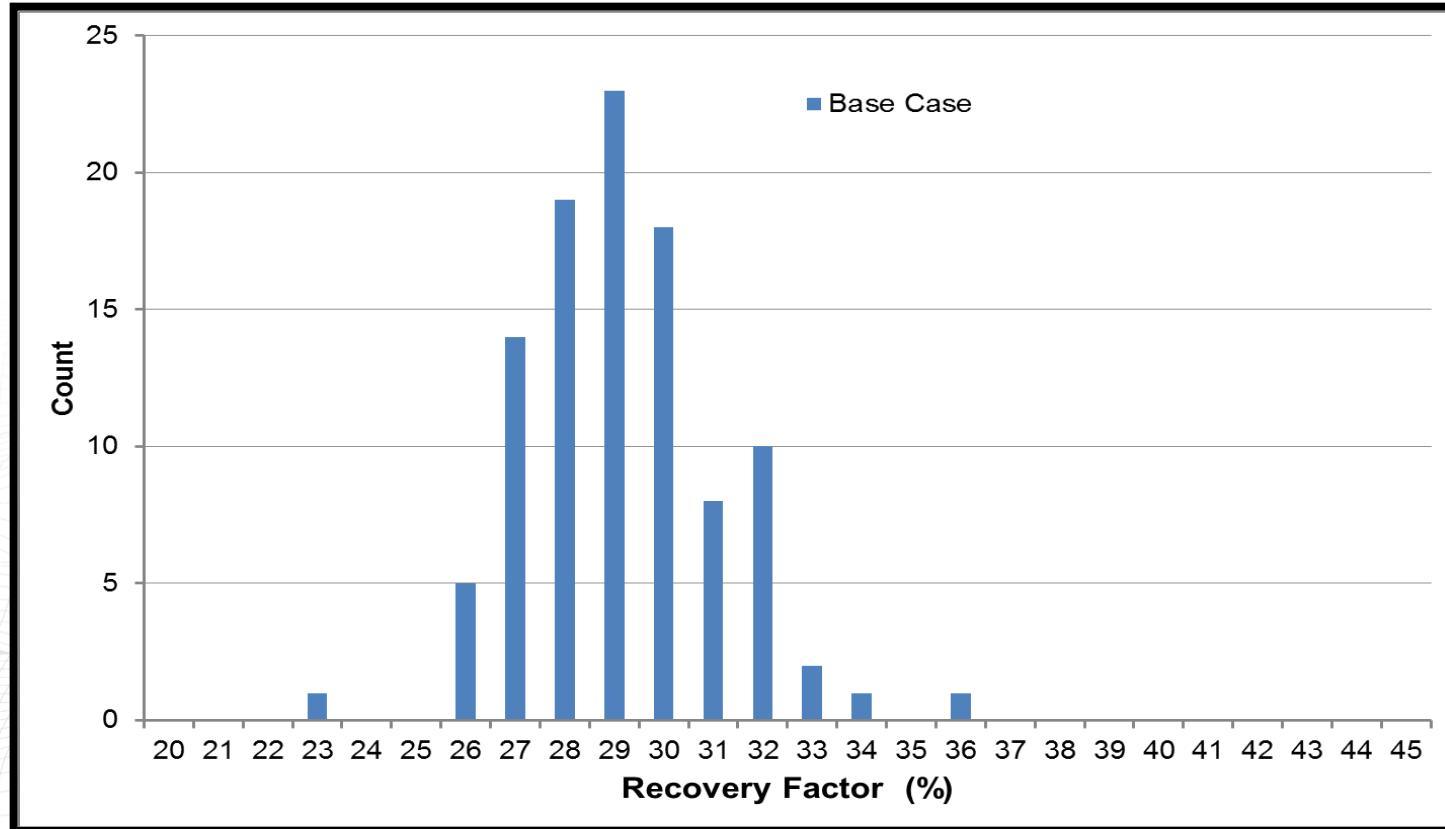
Realization 5



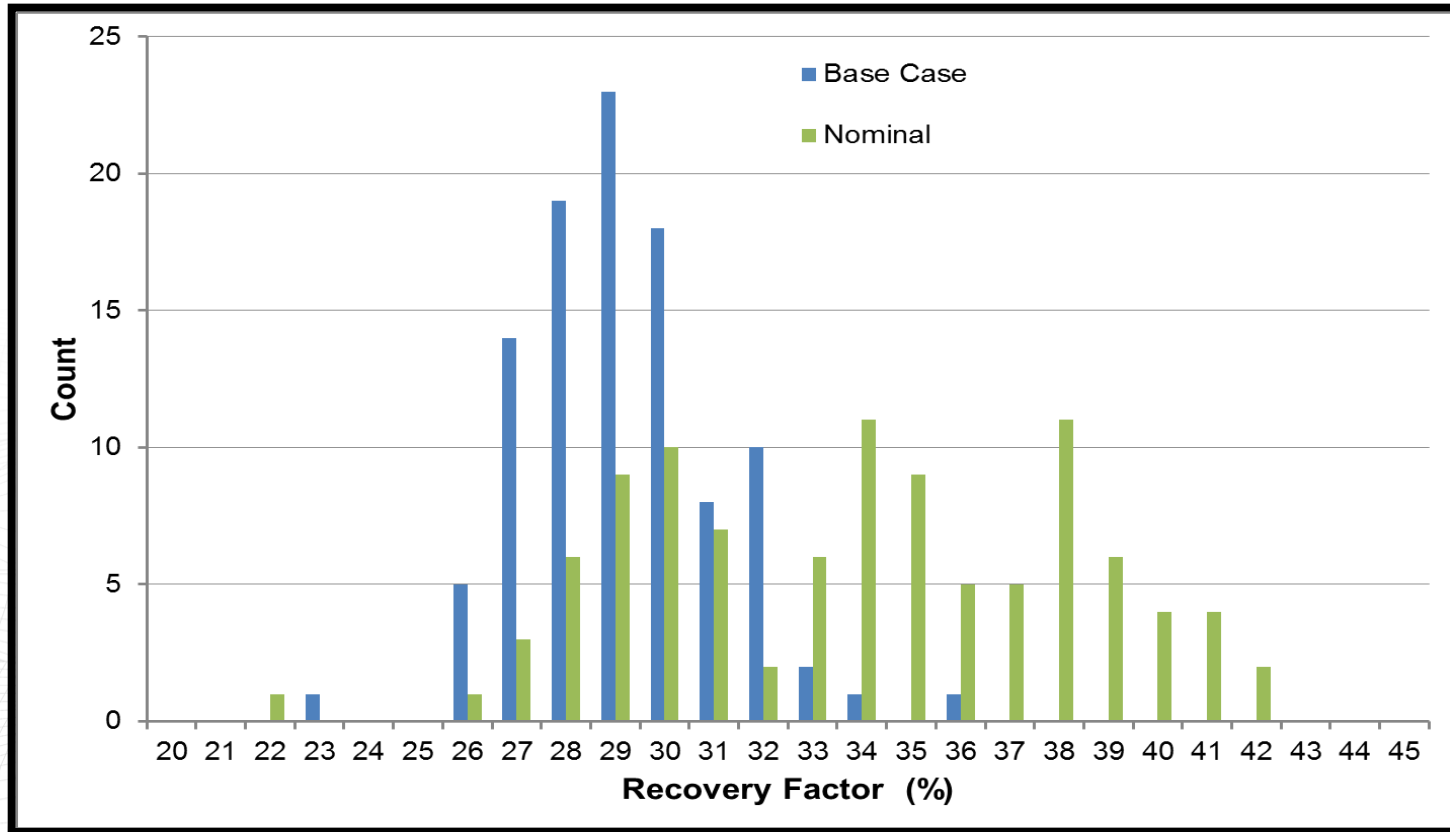
Robust Optimization



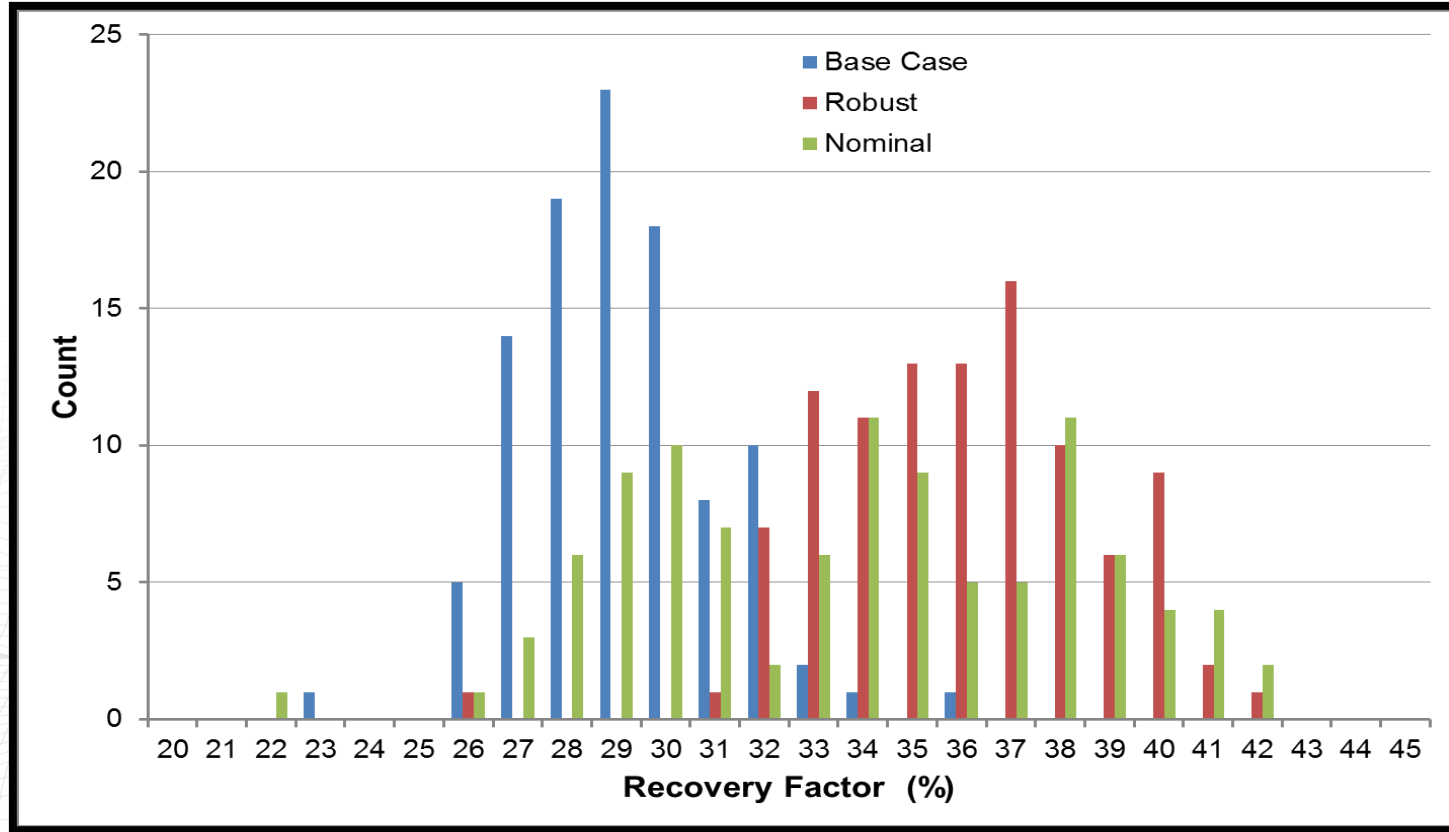
Base Case Histogram



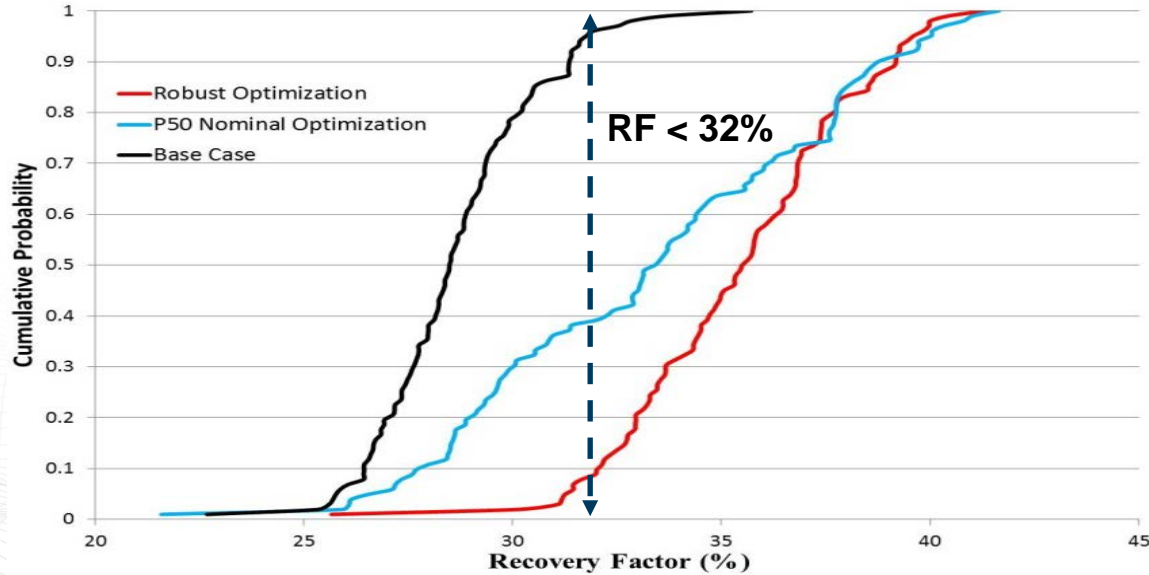
Base Case vs. Nominal



Base Case vs. Nominal & Robust Optimizations



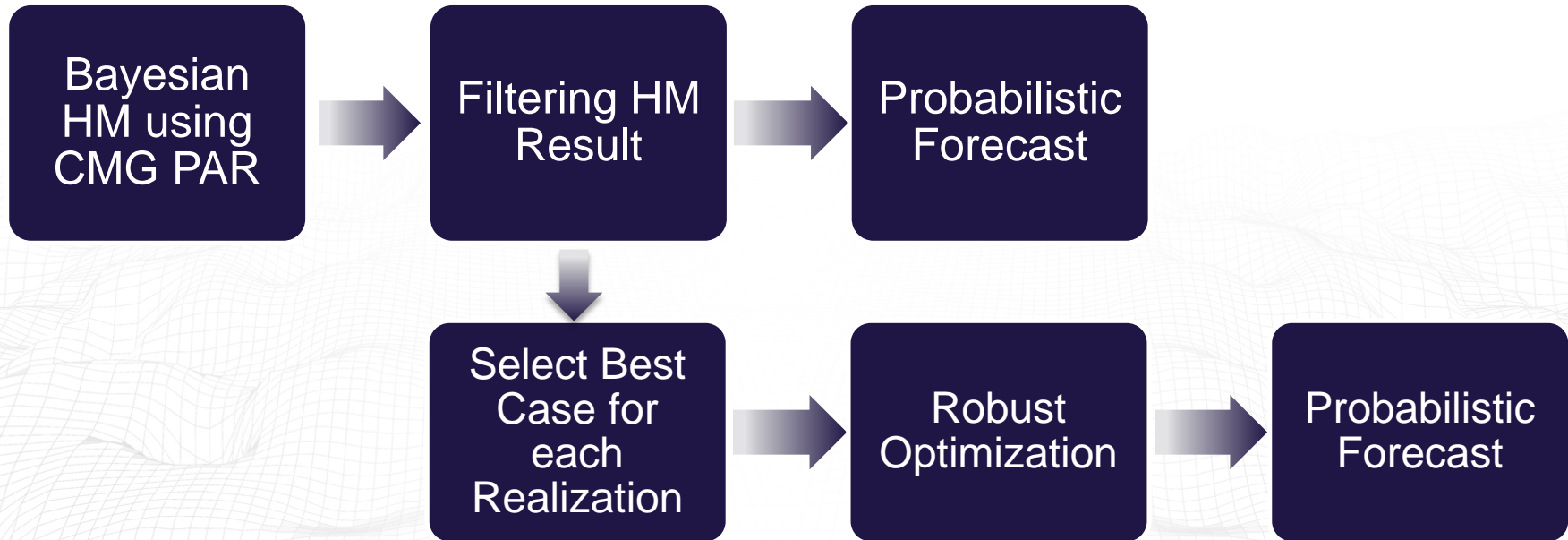
Cumulative Probability: Base Case – Nominal – Robust Optimization



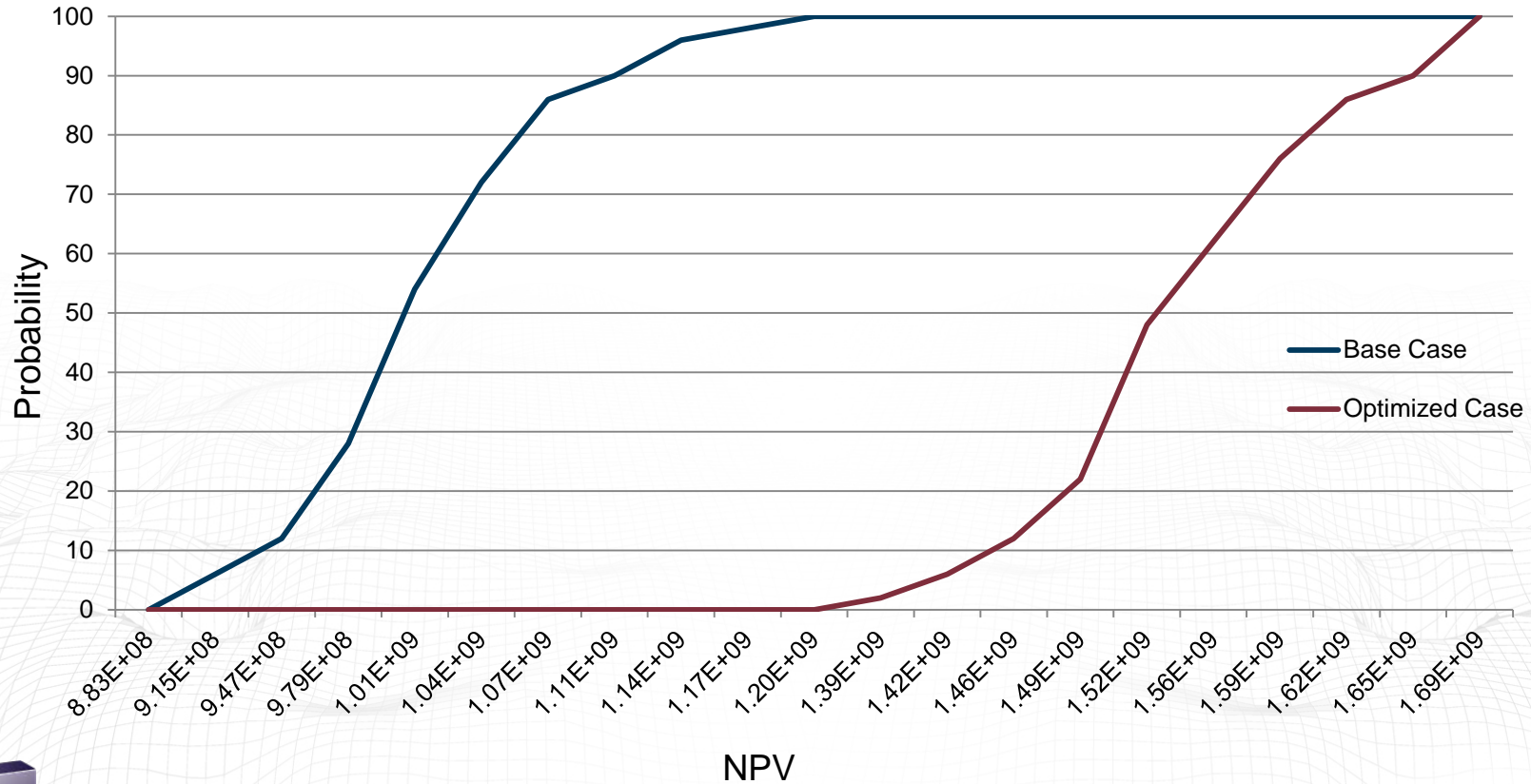
RF < 32%	Probability
Base case	96%
Nominal optimization	39%
Robust optimization	9%



Robust Optimization for Brownfields

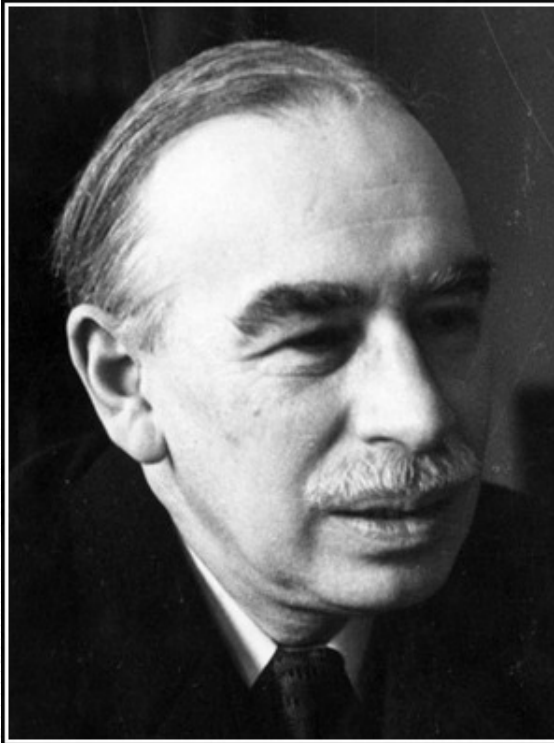


Cumulative Probability: Base Case – Nominal – Robust Optimization



Conclusions

- In today's market more than ever quantifying uncertainty is crucial
- CMG has tools to help carry geologic uncertainty easily through your workflows
- Carrying this uncertainty through the workflow provides immense value in making better decisions



I'd rather be vaguely right than
precisely wrong.

— *John Maynard Keynes* —

AZ QUOTES

CMG's Vision

To be the leading developer and supplier of dynamic reservoir technologies in the **WORLD**