

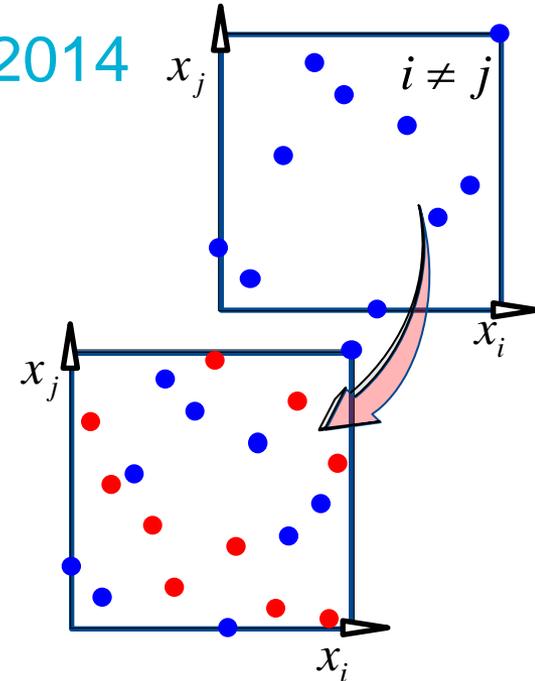
# A Flexible Strategy for Augmenting Design Points For Computer Experiments

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# Starting Point – An industrial Challenge

Let  $f: R^d \rightarrow R$  be a costly to evaluate physical system to be modelled in short time.

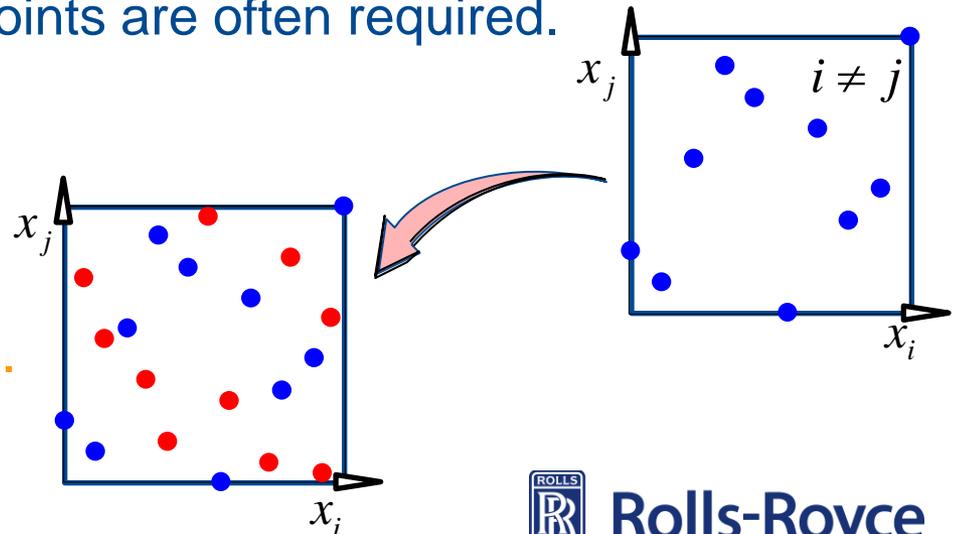
Common Solution: Run a DoE.

**Goal:** Invest as few as possible design evaluations  $N$  to obtain a sufficient understanding, i.e.  $\min_N N$  s.t.  $|f(\mathbf{x}) - \hat{f}(\mathbf{x})| \leq \varepsilon$

*emulated understanding* — *error*

**Problem:** In practice, additional points are often required.

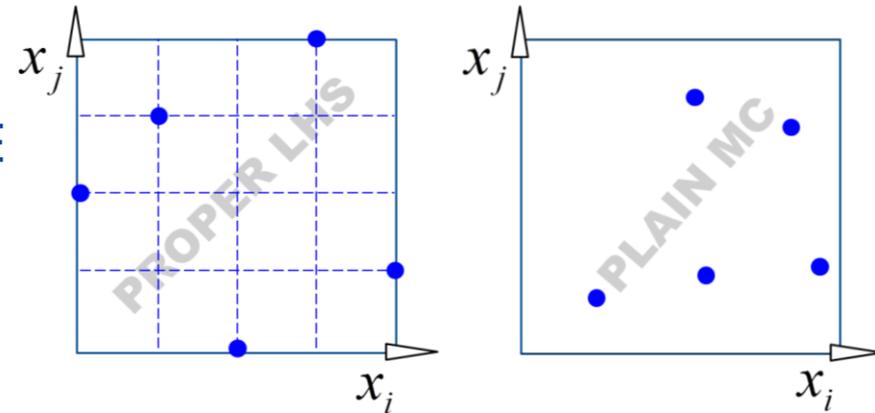
Construct an *optimal* augmented design plan.



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Now, from an industrial perspective an augmentation algorithm needs to

- (i) be **UNIVERSAL**, i.e.  
 - independent from type of initial DoE



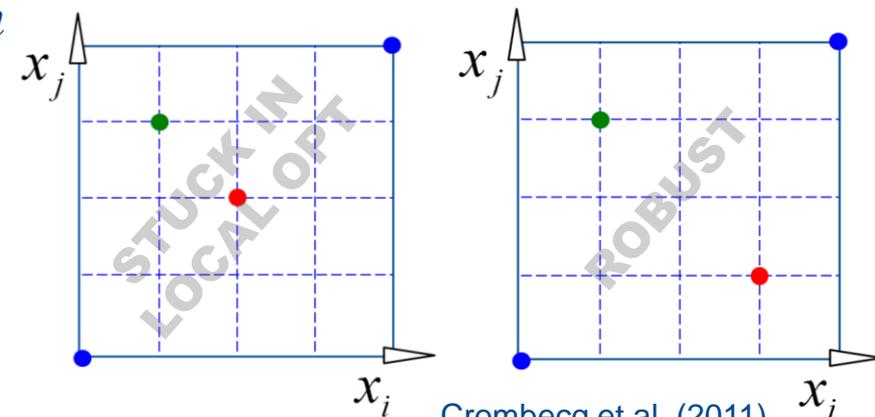
- independent from selected augmentation sequence, i.e.  
 number of levels  $l$  and batch size  $m$

Example with  $l = 2$  and  $m = 1$ :

$N_0 = 2$  (initial points)

$N_1 = 3$  (first augmentation)

$N_2 = 4$  (second augmentation)



Crombecq et al. (2011)

- not restricted to a specific range of design space dimension  $d$   
 or final sample plan size  $N = N_0 + \sum_{i=1}^l m_i$



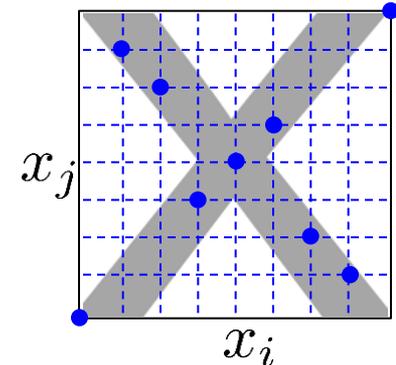
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(ii) improve **SPACE FILLING**, i.e.

- focus on exploration because no model available at the beginning
- evenly spread of design points in design space (avoid X-problem)
- related to projection and orthogonal properties

(iii) be **HIHGLY EFFICIENT**

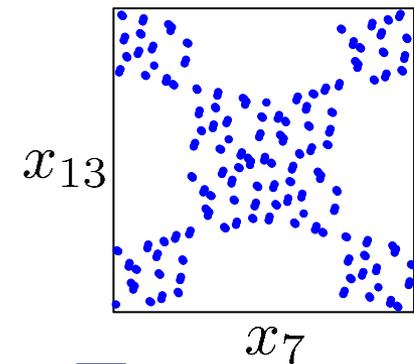
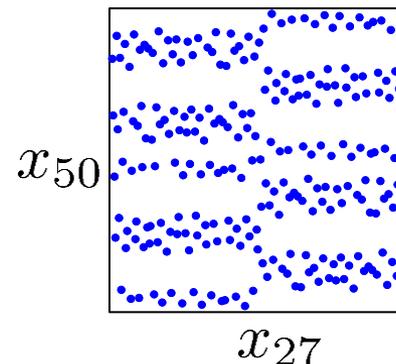
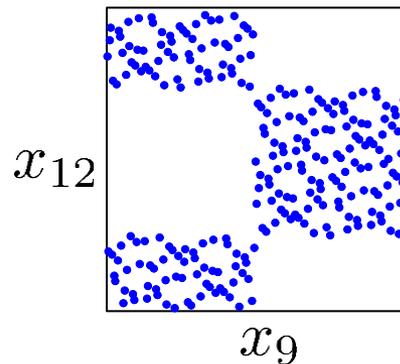
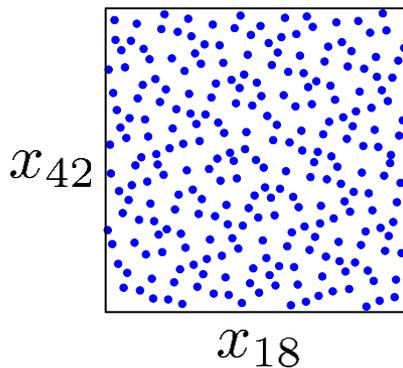
- even for  $d > 50$
- w.r.t. computational time  $t$



Hernandez et al. (2012)

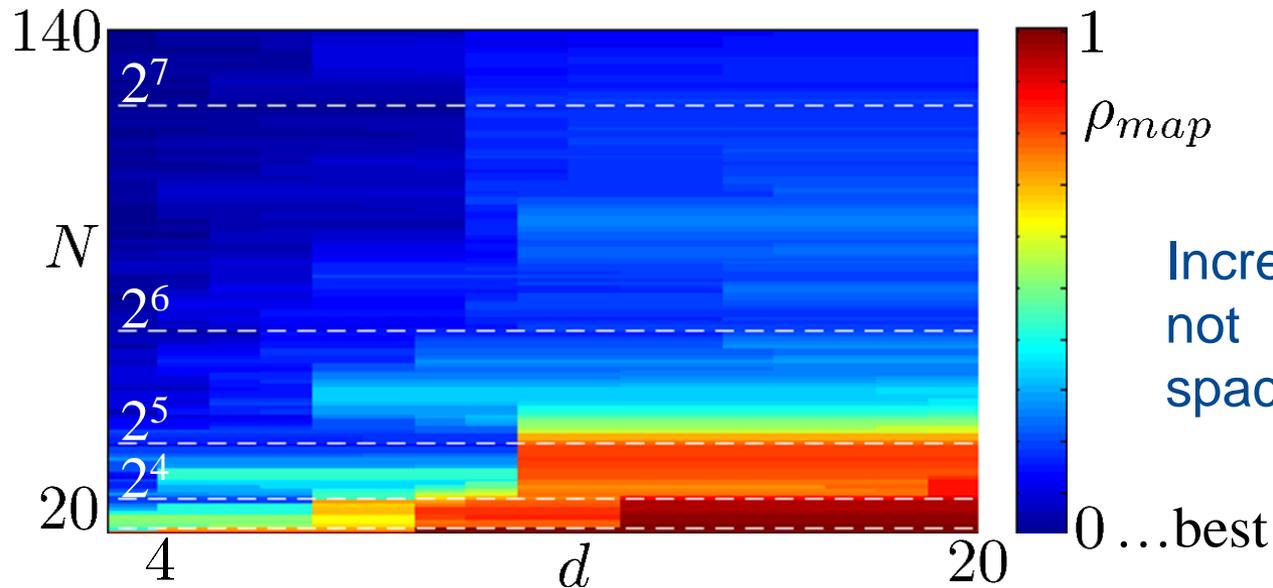
Why e.g. Sobol could fail?

Do some random illustrations for  $d = 50$   $N = 256$



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Evaluate maximum absolute pairwise (map) correlation coefficient defined as  $\rho_{map} = \max_{1 \leq (i,j) \leq d, i \neq j} |\rho_{ij}^P|$  for different  $d$  and  $N$



Can we do better?

## Outline

- Augmentation of (i) an integer grid and (ii) a general grid
- *Optimal* space filling by brute force approach
- Comparative examples



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# Augmentation on Integer or General Grid?

Let  $X^{(k)}$  be the  $k$ -th sample of a DoE,  $k=1(1)N$  and  $\mathbf{x}=[x_1, \dots, x_d]^T \in [0,1]^d$ , to be augmented by  $m$  points.



Classify type of input design plan by strata widths defined as

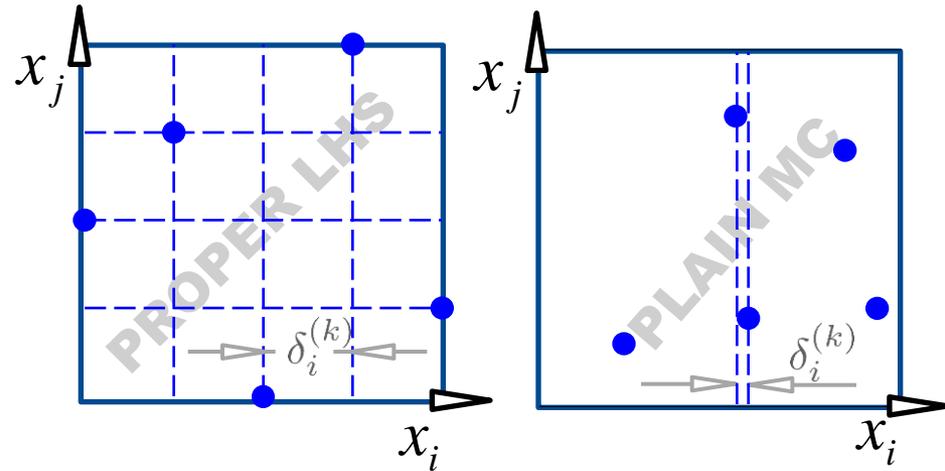
$$\delta_i^{(k)} = x_i^{(k+1:N)} - x_i^{(k:N)} \quad \text{where} \quad x_i^{(1:N)} < x_i^{(2:N)} < \dots < x_i^{(N:N)}$$

with  $k = 1(1)N - 1$  and  $i = 1(1)d$



Compute decision criterion  $\delta = \delta_i^{(k)} - \left\lfloor \frac{\delta_i^{(k)}}{\min_k \delta_i^{(k)}} \right\rfloor \min_k \delta_i^{(k)}$

Do integer grid augmentation if  $\delta \leq 10^{-6}$  and  
general grid augmentation if  $\delta > 10^{-6}$ .



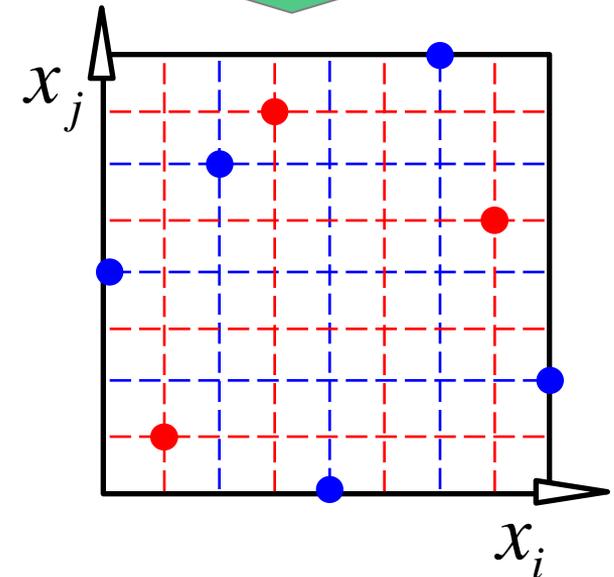
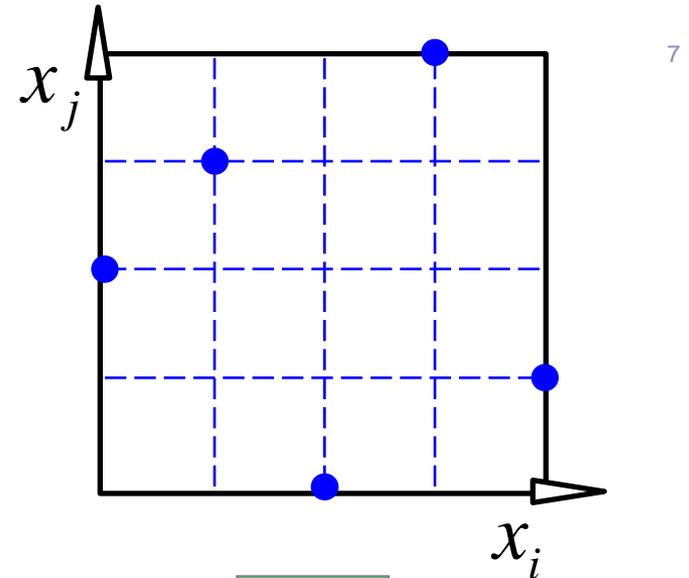
# Integer Grid Augmentation

Check extensibility of the grid by

$$\left\lceil \max_k \frac{1}{\delta_i^{(k)}} + \frac{1}{2} \right\rceil + 1 < N + m$$

If necessary refine the grid. Do identification and indexing of all possible positions for new points.

Generate sets of  $d$  random perturbations  $\{1, 2, \dots, 1/(\delta^{\min} - 1) - N\}$  repeatedly & select set which provides best space-filling criterion.



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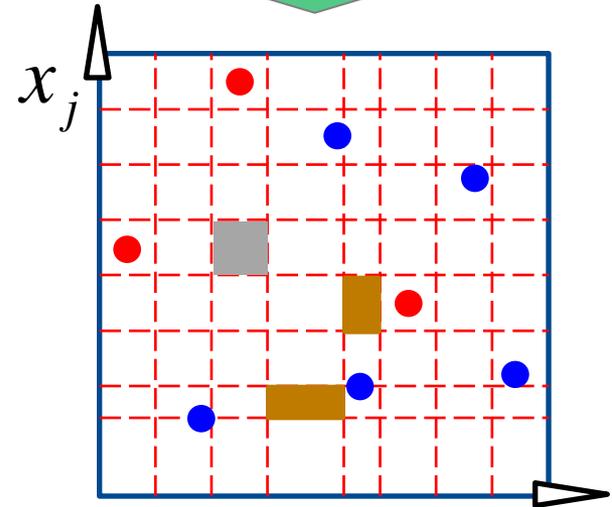
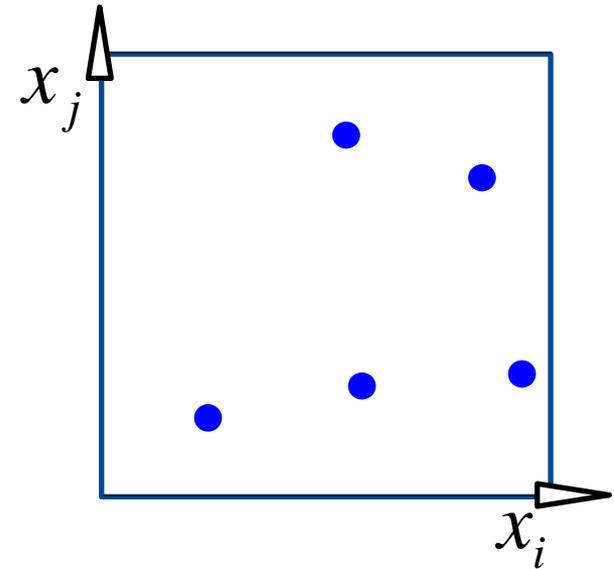
# General Grid Augmentation

Divide dimensions in  $N + m$  equally distributed strata.

Identification of current strata positions for all designs.

From **HYPERCUBES** to **HYPER-CUBOIDS**: Adjust strata bounds till each stratum is used by maximal one design point only.

Generate sets of  $d$  random perturbations  $\{1, 2, \dots, m\}$  repeatedly and select set according to best space-filling criterion to fill empty positions.



# Optimal Space-filling by Brute Force Approach<sup>9</sup>

## How to access quality of a design plan?

- maximum absolute pairwise (map) correlation coefficient (to be minimal)

$$\rho_{map} = \max_{1 \leq (i,j) \leq d, i \neq j} |\rho_{ij}^P|$$

- modified  $L_2$  discrepancy (to be minimal)

$$ML_2 = \left(\frac{4}{3}\right)^d - \frac{2^{1-d}}{N} \sum_{i=1}^N \prod_{j=1}^d (3 - x_{ij}^2) + \frac{1}{N^2} \sum_{i=1}^N \sum_{j=1}^N \prod_{l=1}^d (2 - \max\{x_{ij}, x_{jl}\})$$

Crombecq et al. (2011),  
Joseph and Hung (2008)



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## How to optimize?

- objective function is nonlinear, not differentiable everywhere
- design parameters are natural numbers, i.e. integers
- lattice construct has  $N!^d$  possible Lhs
- for some combinations of  $N$  and  $d$  *complex* algorithms exist

 Let use a brute force approach.

Generate repeatedly permutations for a certain amount of time and select best set of permutations which provides best quality.

 Is this compatible?

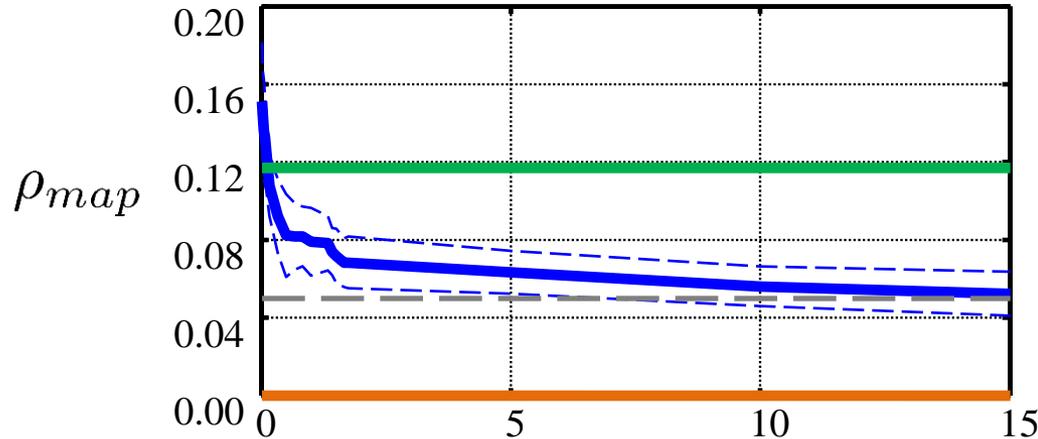
Comparison with results published by Hernandez et al. (2012) and Joseph and Hung (2008) for  $d = 4$  and  $N = 9$  obtained by complex algorithms.



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## 1<sup>st</sup> Scenario

- brute force by minimizing of  $\rho_{map}$  only
- vary improving time  $t_{imp}$  from one second to 15 minutes
- do 30 repetitions for each time do enable statistical statements

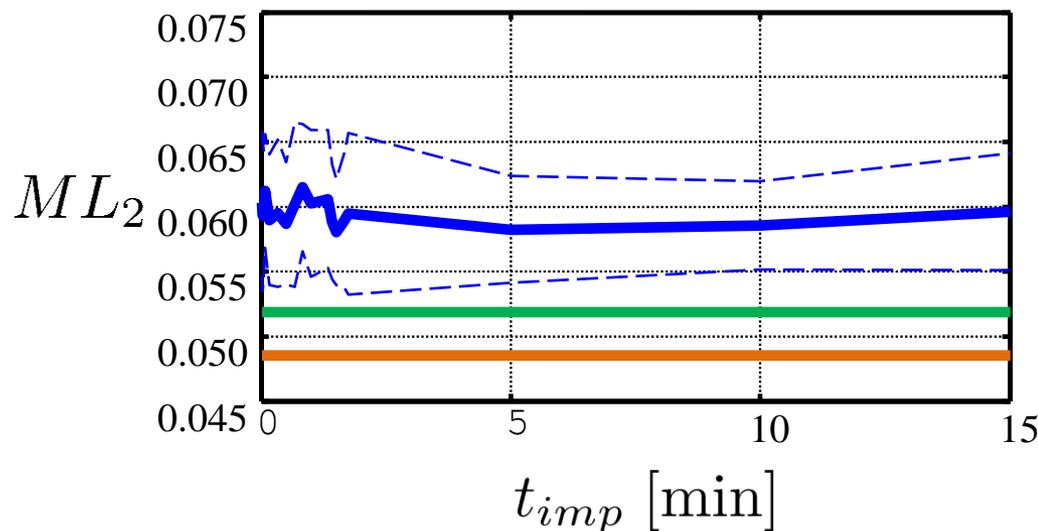


Joseph and Hung (2008)

brute force (mean and +/-std)

nearly orthogonal Lhs

Hernandez et al. (2012), 15 min



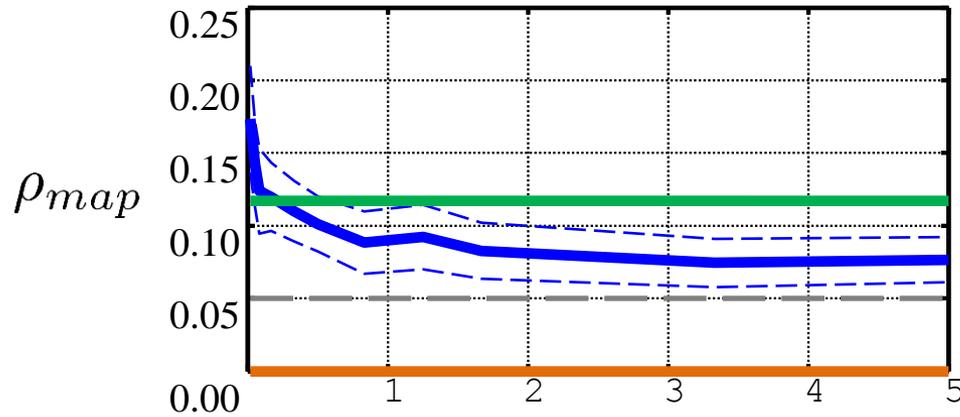
$ML_2$  seems to be  
contradicting to  $\rho_{map}$ .



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## 2<sup>nd</sup> Scenario

- brute force by minimizing of  $0.2\rho_{map} + 0.8ML_2$
- vary improving time  $t_{imp}$  from one second to five minutes
- do 30 repetitions for each time do enable statistical statements

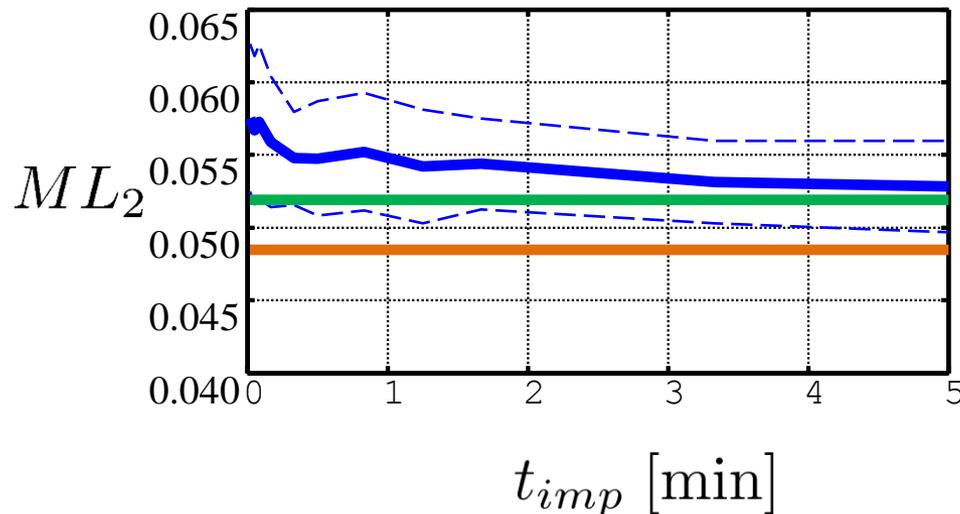


Joseph and Hung (2008)

brute force (mean and +/-std)

nearly orthogonal Lhs

Hernandez et al. (2012), 15 min



Proper weighting coefficients  
lead to a desired trade off.

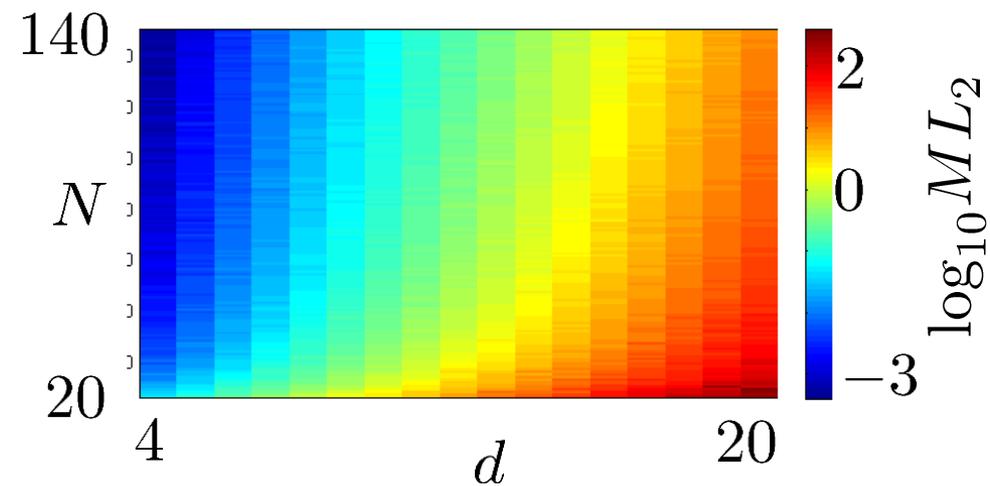
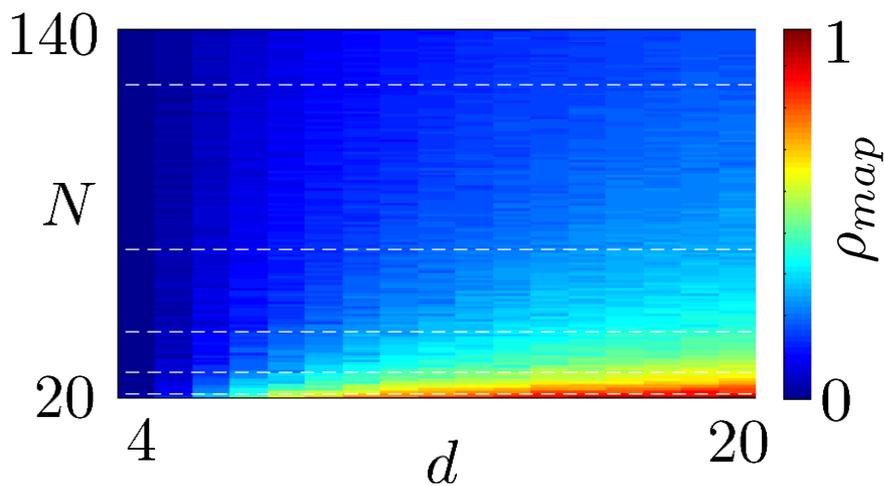
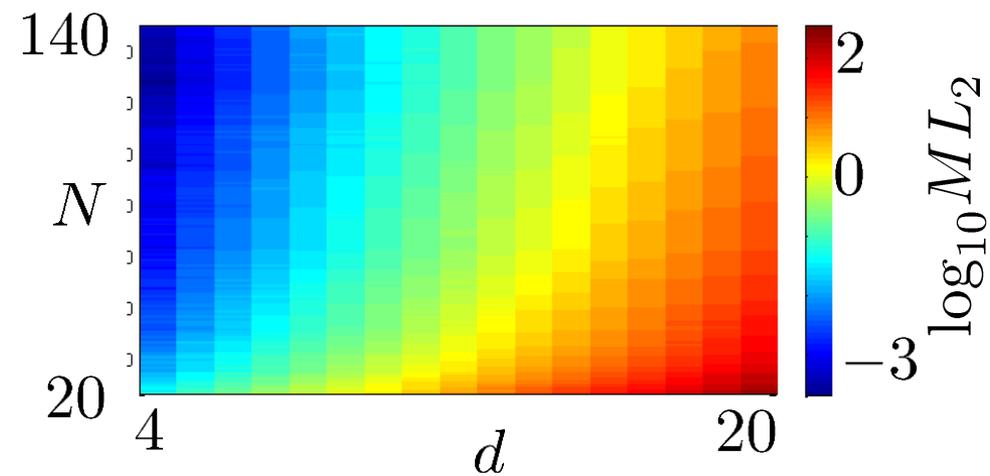
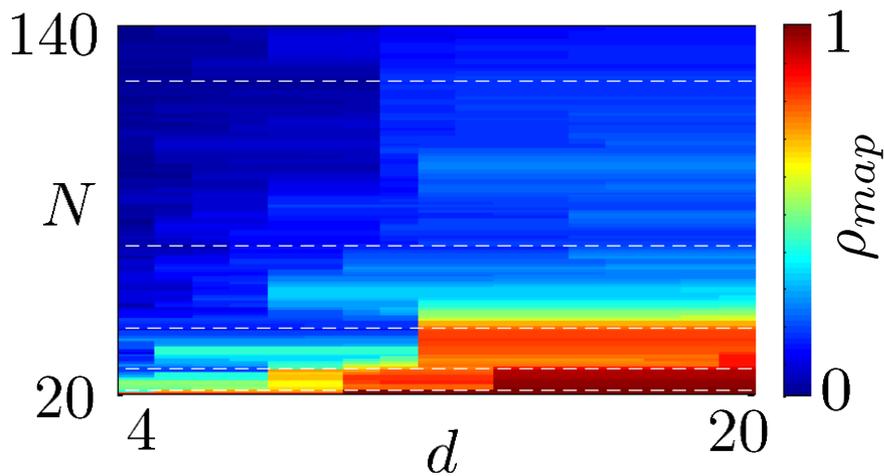


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# A First Comparative Example

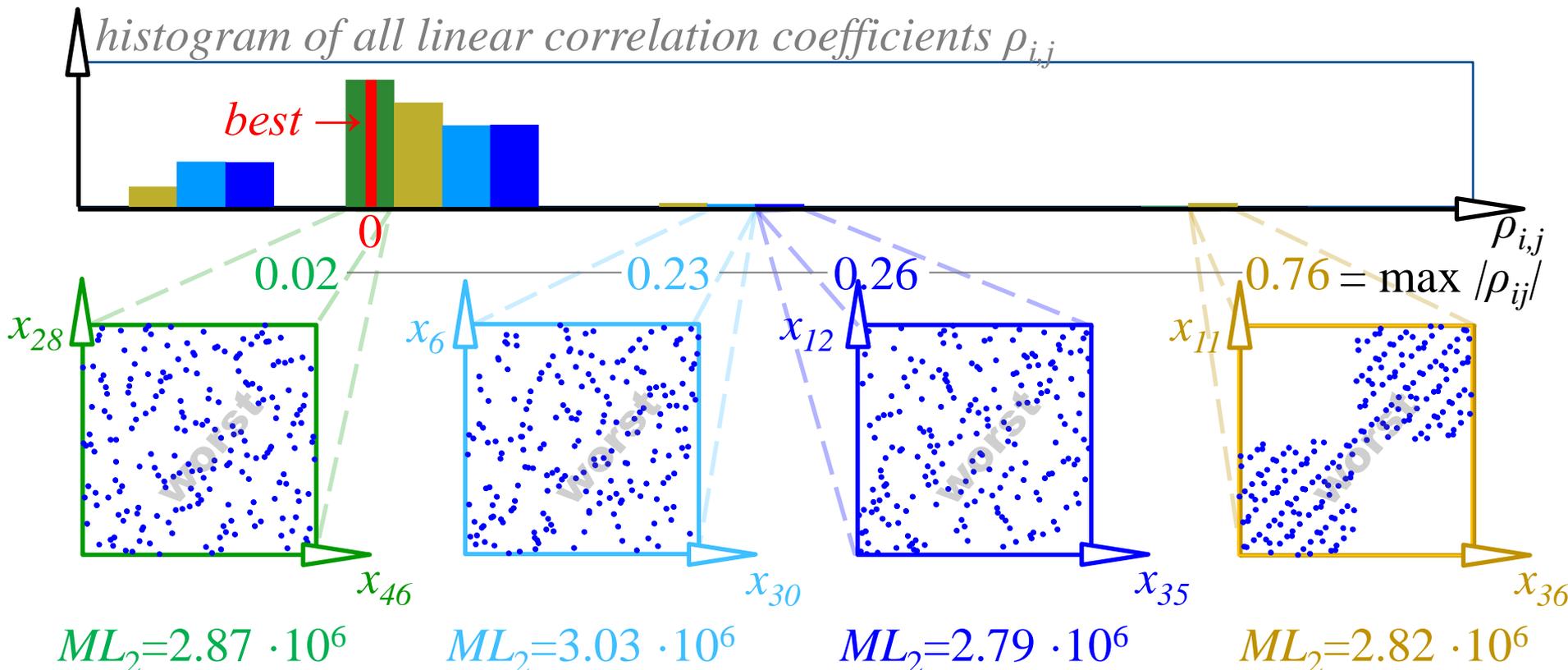
Compare a SOBOL (top) generator with brute force improved Lhs (bottom)

$d \in [3, 20]$   $N \in [5, 150]$   $t_{imp} = 10 \text{ sec}$   $17 \times 145 \text{ grid}$



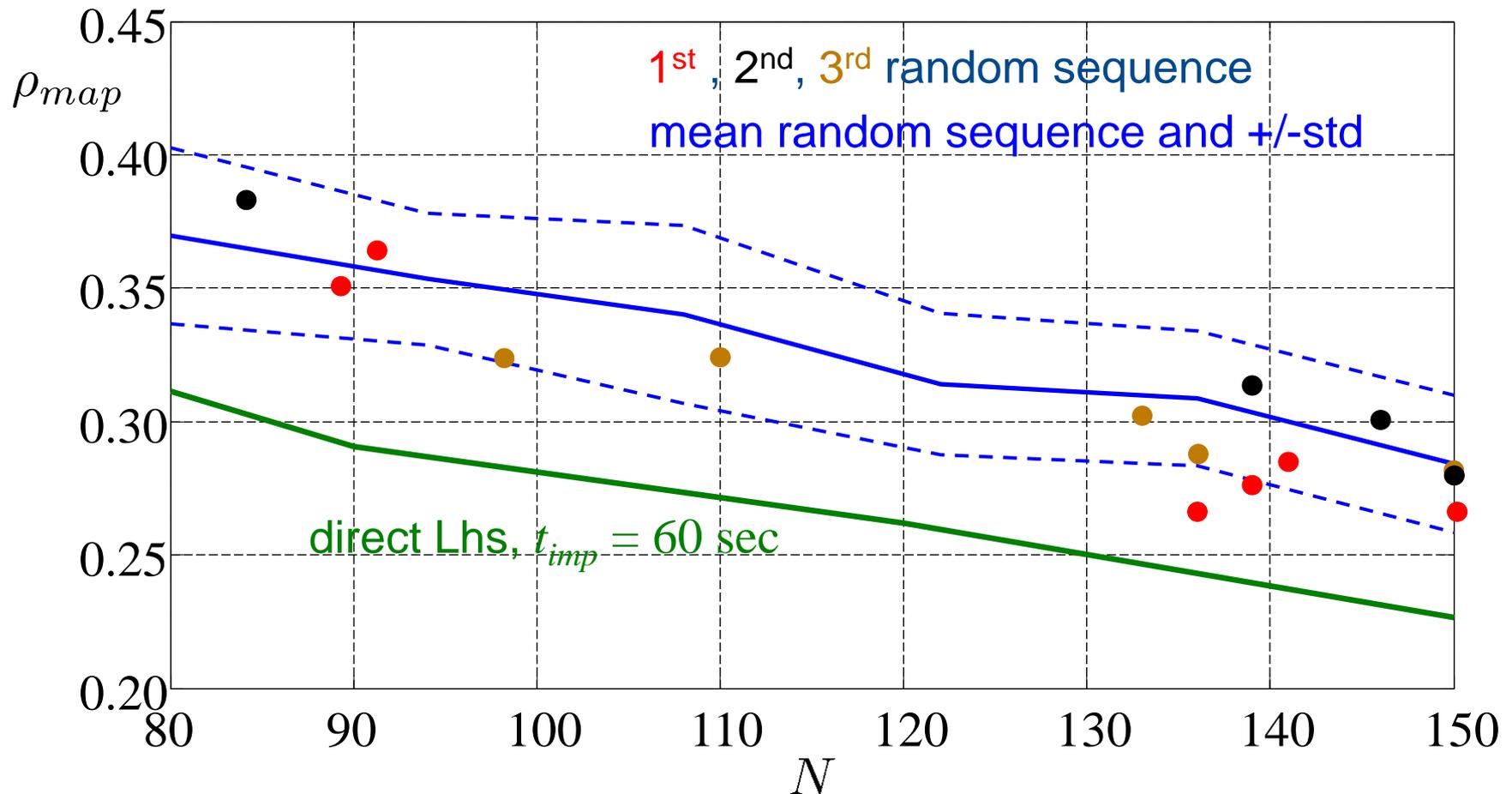
# A Second Comparative Example

Let  $d = 50$ ,  $N_0 = 60$ ,  $N_i = N_{i-1} + m_i$  with  $\mathbf{m} = [40, 50, 50]^T$  be an augmentation sequence applied to an initial **optimal Lhs** (6.6min) and an initial **plain MC** (4.1min). Compare augmented plans with one **optimal Lhs design plan** (20min) and one **SOBOL design plan** (0.1min) with  $N=200$ .



# A Third Comparative Example

Do repeatedly random augmentation sequences (100 times) of an initial Lhs design plan for  $d = 50$ ,  $N_0 = 64$ ,  $N_l = 150$  and  $t_{imp} = 10$  sec, i.e. generate randomly number of levels  $l$  and number of points  $m_i$  to be added at each level.



# Summary and Outlook

A universal algorithm for sequentially augmenting computer experiments is presented. Preliminary results show potentials related to space-filling even for design space dimensions of  $d = 50$ .

Test proposed method against other quasi-random low-discrepancy sequences.

Implement a capability to add a factor/design parameter instead of augmenting number of samples.

## References

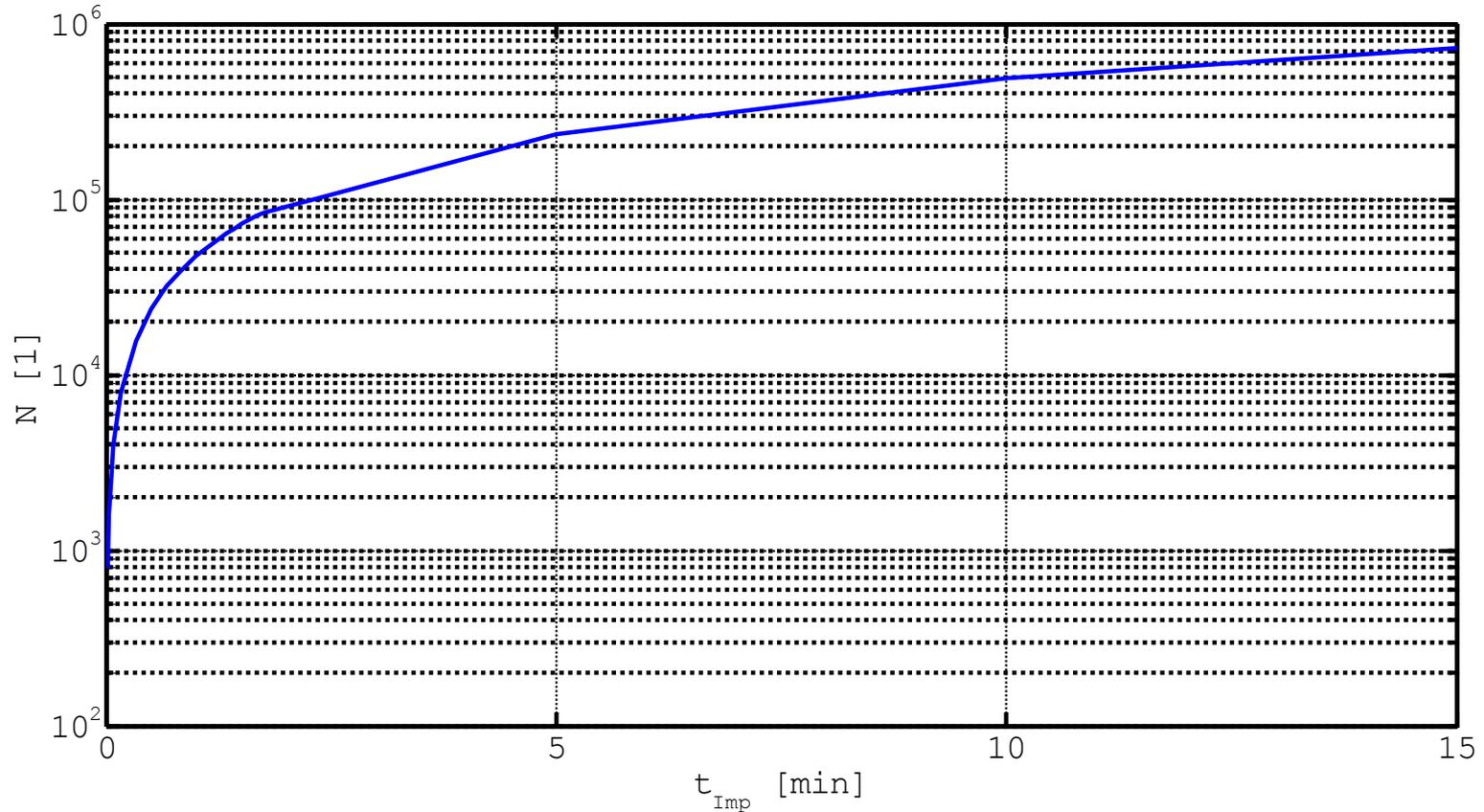
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# Back Up Slides

Average number of generated Lhs design plans for 1<sup>st</sup> Scenario.



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