Optimization Techniques

Laboratory 9

Design of Experiments

cc.rambaldimigliore@unitn.it elia.cunegatti@unitn.it mvincze@fbk.eu

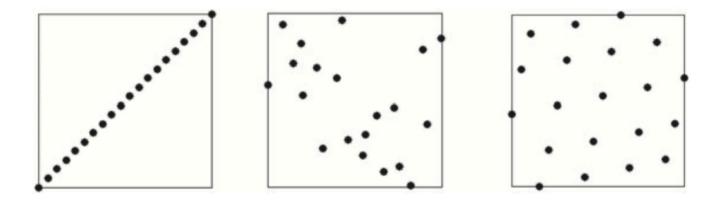


UNIVERSITY OF TRENTO - Italy

Information Engineering and Computer Science Department

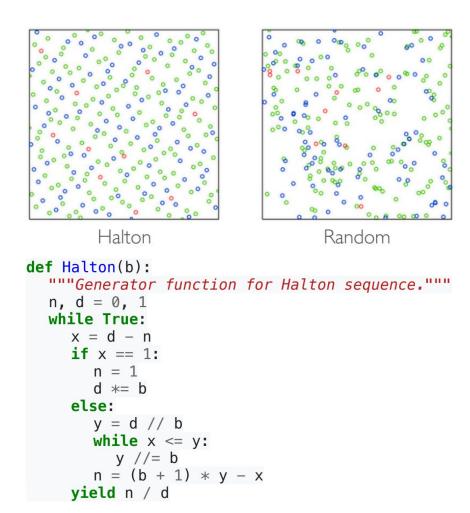
DOEs

- Space-filling sampling methods
- Initialization step to many optimization methods
- Can be used to get insight about the search space



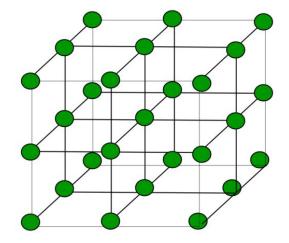
Halton

- Generate samples in a normalized space
- Deterministic sample construction based on coprimes
- Quasi-random low-discrepancy
- Performs well in low dimensions



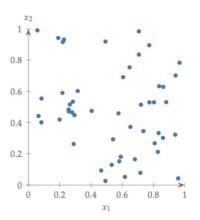
Full Factorial

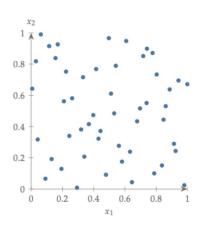
- N variables + M divisions / variable == NxM combinations
- Only applicable for low number of variables



Latin Hypercube

- Each sample is the only one in each axis-aligned hyperplane containing it
- Ensures that each sampling space dimension is roughly evenly sampled





Inputs:

 n_s : Number of samples n_d : Number of dimensions $P=\{P_1,\ldots,P_{n_d}\}$: (optionally) A set of cumulative distribution functions

Outputs:

$$X = \{x_1, \dots, x_{n_s}\}$$
: Set of sample points

for
$$j=1$$
 to n_d do for $i=1$ to n_s do
$$V_{ij}=\frac{i}{n_s}-\frac{R_{ij}}{n_s} \text{ where } R_{ij} \in \mathbb{U}[0,1] \text{ Randomly choose a value in each equally spaced cell from uniform distribution}$$

end for

$$X_{*j} = P_j^{-1}(V_{*j})$$
 where P_j is a CDF Evaluate inverse CDF Randomly permute the entries of this column X_{*j} Alternatively, permute the indices $1 \dots n_s$ in the prior for loop

end for