

# Emulators and Design

Andrew Parnell and Philip Cardiff  
andrew.parnell@mu.ie



[https://andrewcparnell.github.io/intro\\_emulators/](https://andrewcparnell.github.io/intro_emulators/)

# Introduction

- ▶ In this part of the course we will cover how to choose the input values at which to run our simulator
- ▶ Recall that the simulator is slow to run and we can only afford a small number of runs
- ▶ We need to choose the 'best possible' input values to run the simulator
- ▶ The values that we choose will become the inputs (features) to our machine learning model; the outputs will become the targets
- ▶ How do we choose these best values?

## Choosing the total number of runs of the simulator

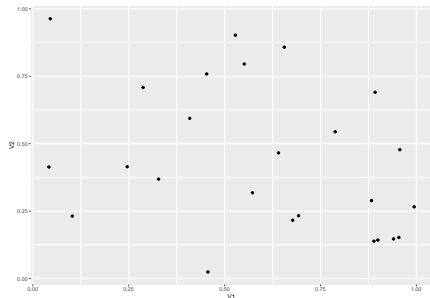
- ▶ Choosing the total number of runs is not really a mathematical/statistical problem, but it needs to be as large as possible
- ▶ It depends on the speed of the simulator, and how much patience/super computer time we have to spend on running it
- ▶ With some assumptions about the variability of the simulator surface it is possible to work out the expected uncertainty in the emulator for a given number of runs, but in the end it always comes down to doing as many as you possibly can

Choosing which input values to run the simulator at is a much harder problem. Here are some ideas...

## Bad idea number 1 - random values

We could just choose random values across the input space:

```
N_sim <- 25  
N_input <- 2  
X <- matrix(runif(N_sim * N_input), ncol = N_input, nrow = N_sim)
```

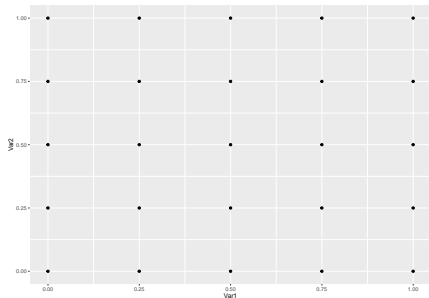


This is bad because we might miss large chunks of the input space; and it gets worse in higher dimensions

## Bad idea number 2 - grids

Alternatively we could grid up the input space to cover the full region

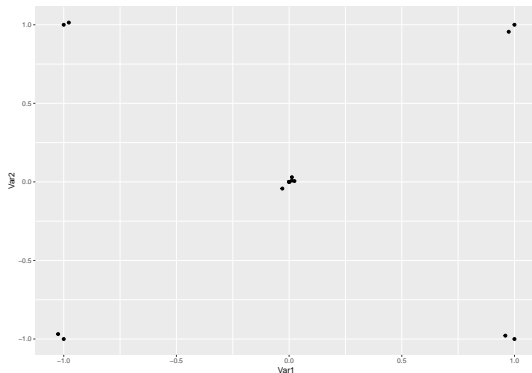
```
x <- seq(0, 1, length = sqrt(N_sim))  
X <- expand.grid(x, x)
```



... but this is also bad because (a) one or more of the variables might not be important (so harder to identify non-linear effects)

## Bad idea number 3 - traditional design

Here's a Central-Composite (similar to Box-Behnken) design

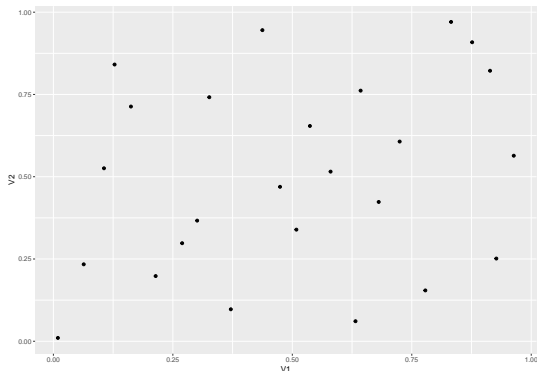


... but since the simulator is deterministic there is no point in running it at the same input points more than once

## A better idea - Latin hypercubes

A better method is a Latin hypercube design

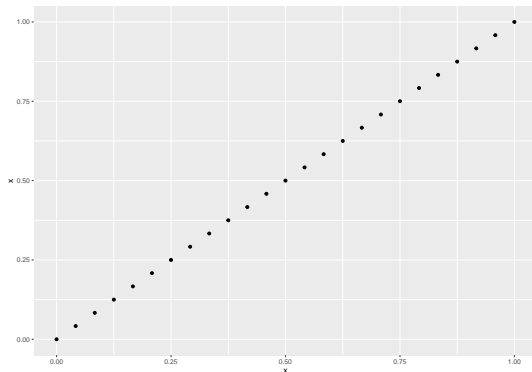
```
library(lhs)  
X <- maximinLHS(N_sim, N_input)
```



Think of dividing up the input space into horizontal and vertical bands, and picking one value that covers each row and each column

## Bad Latin hypercubes

Actually, that idea doesn't work very well, because this is also a valid Latin hypercube design:



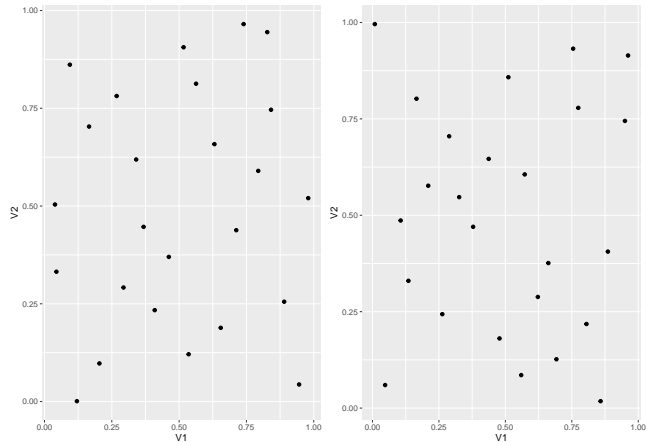
We want a design that is both a Latin hypercube and fills as much of the space as possible



## Good Latin hypercubes

Try to optimise the sample by finding a Latin Hypercube sample that maximises the minimum distance between design points

We can generate lots of these with the function `maximinLHS`:



## Designs for example 1

- Recall example 1: our simple sine wave with 2 inputs:

```
f <- function(x1, x2) {  
  return(10 * sin(pi * x1 * x2))  
}
```

Suppose we are willing to run this simulator 20 times. Create the design with:

```
n_runs <- 20  
n_inputs <- 2  
initial_grid <- maximinLHS(n_runs, n_inputs)
```

## Designs for example 2

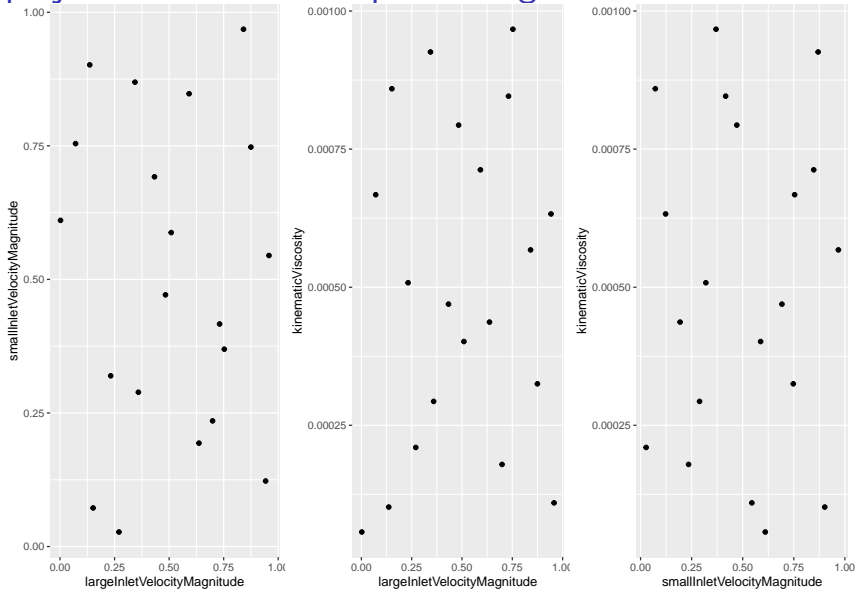
Recall example 2 is our 2D Navier Stokes model with 3 inputs

Suppose we are willing to run this 20 times:

```
n_runs <- 20  
n_inputs <- 3  
initial_grid <- maximinLHS(n_runs, n_inputs)
```

... harder to plot this as now in 3 dimensions

## Some 2D projections of the example 2 design



## Where to get more information on design

There are many possibilities when it comes to design in emulation:

- ▶ More advanced ways of spreading out the design and ensuring you do not miss important parts of the space
- ▶ More advanced ways of taking into account knowledge about the likely values of the inputs. For example, we might be able to guess a probability distribution for the input variables and use this to 'target' values in the design
- ▶ See Chapter 5 of [The Design and Analysis of Computer Experiments](#) for more detailed discussion
- ▶ We will stick to using Maximin Latin hypercube samples for our emulator

# Summary

- ▶ We need to choose how many total runs we can afford
- ▶ Lots of traditional design ideas do not work well for emulator design
- ▶ Remember that the simulator is deterministic - not much point running at the same values twice! And try to avoid gridded values if you can
- ▶ Using Maximin Latin hypercube samples a good default choice
- ▶ Next: building the emulator...