5 designs and 5 iterations and 3 granularities all of whose categories change

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1 *Simulation design*

```
t = 300
n <- seq(0, t, 1)
g1 <- n %%2
g2 <- n %%3
g3 <- n %%5
mu1 = c(0, 0)
mu21= c(2, 0, 0)
mu22 = c(0, 2, 0)
mu23 = c(0, 0, 2)
mu3 = c(0, 0, 0, 0, 0)
```

Three circular granularities g1, g2 and g3 are considered with levels 2, 3 and 5 respectively. Many time series with 300 observations are created using the five designs below, each of which is iterated five times. We anticipate to have 3 clusters, each with five time series conforming to the same design, once we execute the clustering.

2 *Distance computation*

1) wpd for three granularities computed and then distance between designs are computed by computing the euclidean distances between them - gran as variables, designs across rows and cell values as wpd.

2) Robust scaling done for each customer and then JSD is computed between same categories for a granularity.

3 *Granularities design*

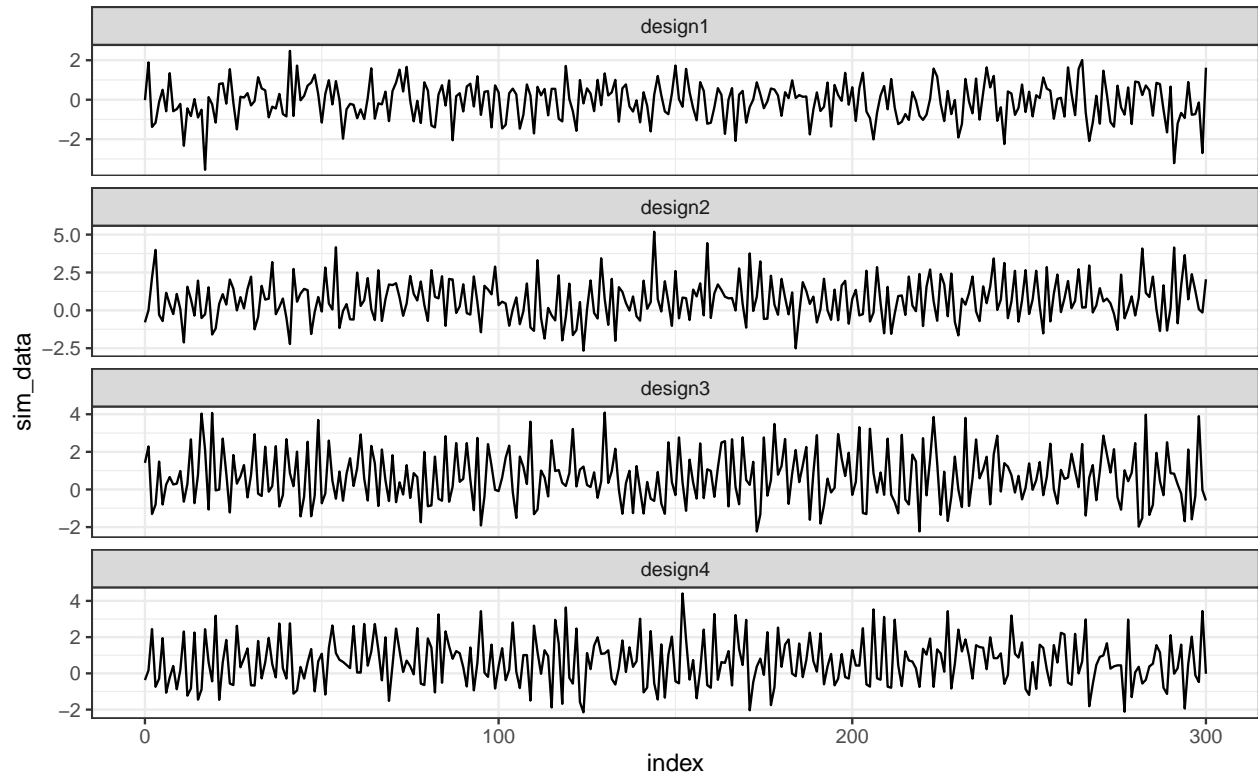## 4 *Simulate data*

```
## # A tibble: 1,204 x 6
##    index    g1    g2    g3 design  sim_data
##    <dbl> <dbl> <dbl> <dbl> <chr>      <dbl>
## 1      0     0     0     0 design1 -0.0218
## 2      0     0     0     0 design2 -0.793
## 3      0     0     0     0 design3  1.41
## 4      0     0     0     0 design4 -0.370
## 5      1     1     1     1 design1  1.89
## 6      1     1     1     1 design2  0.00233
```
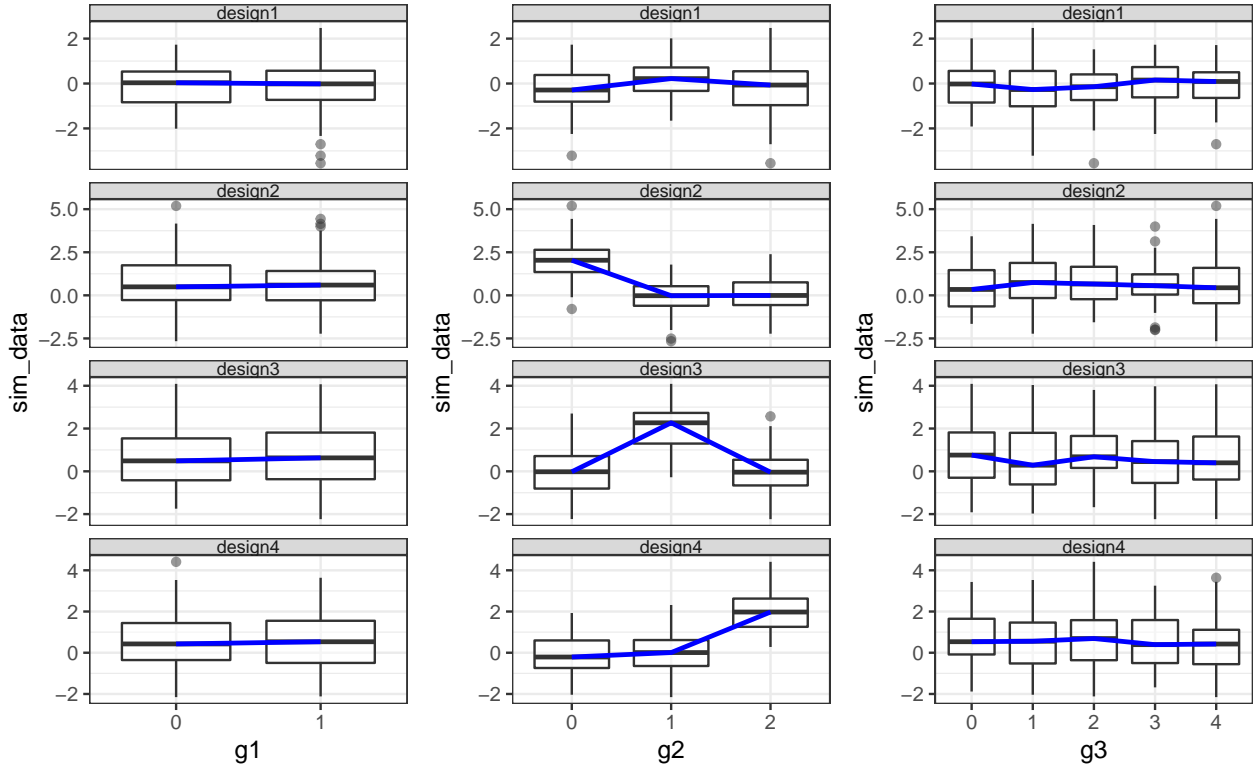
```
## 7     1     1     1     1 design3  2.29
## 8     1     1     1     1 design4  0.169
## 9     2     0     2     2 design1 -1.38
## 10    2     0     2     2 design2  2.16
## # ... with 1,194 more rows
```

5   *Plot raw data*

6 Plot distribution across granularities

7 Iterate designs

```
## # A tsibble: 6,020 x 8 [1]
## # Key:        customer_id [20]
##    seed_id index    g1    g2    g3 design  sim_data customer_id
##    <chr>   <dbl> <dbl> <dbl> <dbl> <chr>      <dbl> <chr>
##  1 1          0     0     0     0 design1    0.586 design1-1
##  2 1          1     1     1     1 design1    0.709 design1-1
##  3 1          2     0     2     2 design1   -0.109 design1-1
##  4 1          3     1     0     3 design1   -0.453 design1-1
##  5 1          4     0     1     4 design1    0.606 design1-1
##  6 1          5     1     2     0 design1   -1.82  design1-1
##  7 1          6     0     0     1 design1    0.630 design1-1
##  8 1          7     1     1     2 design1   -0.276 design1-1
##  9 1          8     0     2     3 design1   -0.284 design1-1
## 10 1          9     1     0     4 design1   -0.919 design1-1
## # ... with 6,010 more rows
```

8 Cluster designs: distance using wpd

```
##            Reference
## Prediction design1 design2 design3 design4
##     design1       5       0       0       0
##     design2       0       2       1       2
##     design3       0       2       1       1
##     design4       0       1       3       2
```

9 Cluster designs: distance using js distance between categories (robust scaling)

```
##            Reference
## Prediction design1 design2 design3 design4
##     design1       5       0       0       0
##     design2       0       5       0       0
##     design3       0       0       5       0
##     design4       0       0       0       5
```

10 Cluster designs: distance using js distance between categories (nqt)

```
##            Reference
## Prediction design1 design2 design3 design4
##     design1       5       0       0       0
##     design2       0       5       0       0
##     design3       0       0       5       0
##     design4       0       0       0       5
```