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

Sayani Gupta

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#### 1 Simulation design

```
t = 300

n <- seq(0, t, 1)

g1 <- n %%2

g2 <- n %%3

g3 <- n %%5

mu11 = c(0, 5)

mu12 = c(0, 3)

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.

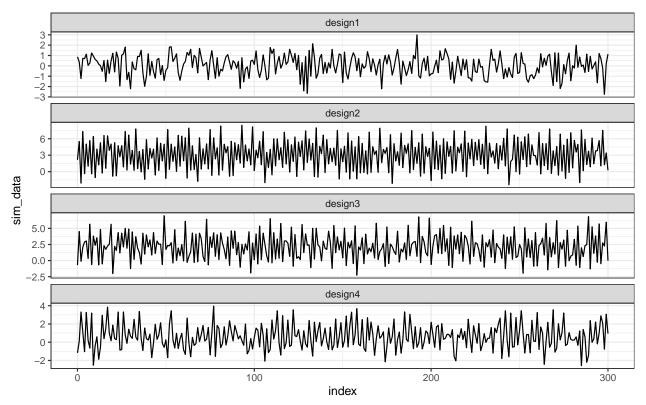
## 3 Granularities design

#### 4 Simulate data

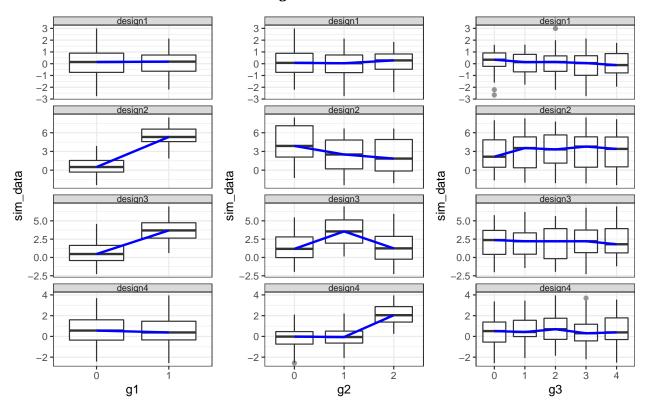
```
# A tibble: 1,204 x 6
##
##
      index
                g1
                       g2
                             g3 design
                                         sim_data
##
      <dbl> <dbl> <dbl> <dbl> <chr>
                                             <dbl>
##
    1
           0
                 0
                        0
                               0 design1
                                             0.873
##
    2
           0
                 0
                        0
                               0 design2
                                             2.16
    3
           0
                 0
                        0
##
                               0 design3
                                            -0.704
                 0
                        0
                               0 design4
    4
           0
                                            -1.17
##
    5
                        1
                               1 design1
                                             0.377
```

```
##
          1
                 1
                              1 design2
                                            5.50
                        1
##
    7
          1
                 1
                        1
                              1 design3
                                            4.52
                                            0.113
##
##
    9
          2
                 0
                        2
                              2 design1
                                            -1.23
          2
                        2
                                            -2.09
##
   10
                 0
                              2 design2
##
         with 1,194 more rows
```

#### 5 Plot raw data



#### 6 Plot distribution across granularities



## $7 \quad Iterate \ designs$

```
# A tsibble: 66,220 x 8 [1]
                 customer_id [220]
   # Key:
##
      seed_id index
                        g1
                              g2
                                     g3 design sim_data customer_id
               <dbl> <dbl> <dbl> <dbl> <chr>
                                                    <dbl> <chr>
##
    1 1
                   0
                         0
                                      0 design1
                                                    0.586 design1-1
##
                                0
    2 1
                                      1 design1
##
                   1
                                                    0.709 design1-1
##
    3 1
                   2
                         0
                                2
                                      2 design1
                                                   -0.109 design1-1
                   3
##
                                0
                                      3 design1
                                                   -0.453 design1-1
##
    5 1
                   4
                         0
                                1
                                      4 design1
                                                    0.606 design1-1
                   5
                                2
    6 1
                                      0 design1
                                                   -1.82 design1-1
##
                   6
##
    7 1
                         0
                                0
                                      1 design1
                                                    0.630 design1-1
                   7
                                      2 design1
##
    8 1
                                                   -0.276 design1-1
##
    9 1
                   8
                                2
                                      3 design1
                                                   -0.284 design1-1
                   9
                                      4 design1
                                                   -0.919 design1-1
## # ... with 66,210 more rows
```

## 8 Cluster designs: distance using wpd

##	Reference						
##	${\tt Prediction}$	design1	design2	design3	${\tt design4}$		
##	design1	55	0	0	0		
##	design2	0	55	0	0		
##	design3	0	0	55	0		
##	design4	0	0	0	55		

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

##	I				
##	${\tt Prediction}$	design1	design2	design3	design4
##	design1	55	0	0	0
##	design2	0	55	0	0
##	design3	0	0	55	0
##	design4	0	0	0	55

# $\begin{array}{ll} 10 & \textit{Cluster designs: distance using js distance between categories} \\ & (nqt) \end{array}$

##	F				
##	${\tt Prediction}$	design1	design2	design3	design4
##	design1	55	0	0	0
##	design2	0	55	0	0
##	design3	0	0	55	0
##	design4	0	0	0	55