Choosing an appropriate scalar transformation to normalise wpd

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• Data presented

Observations are generated from a N(0,1) distribution for each combination of nx and nfacet from the following sets: $nx = nfacet = \{2, 3, 5, 7, 14, 20, 31, 50\}$ to cover a wide range of levels from very low to moderately high. Each combination is being referred to as a panel. That is, data is being generated for each of the panels $\{nx = 2, nfacet = 2\}, \{nx = 2, nfacet = 3\}, \{nx = 2, nfacet = 5\}, \dots, \{nx = 50, nfacet = 31\}, \{nx = 50, nfacet = 50\}$. For each of the 64 panels, ntimes = 500 observations are drawn for each combination of the categories. That is, if we consider the panel $\{nx = 2, nfacet = 2\}, 500$ observations are generated for each of the combination of categories from the panel, namely, $\{(1,1), (1,2), (2,1), (2,2)\}$. The values of λ is set to 0.67 and values of raw wpd is obtained.

- How the distribution of the raw wpd (without any transformation for normalization) looks across nfacets and nx? Both shape and scale of the distribution changes for different nx and nfacet categories as could be seen in Figure ??.
- Plot the values of raw wpd against nx*nfacet to see the rough relationship (??).

Current attempt

• Fit a log-linear model of median(values) to log(nx*nfacet) as discussed today. The fit is pretty good and the summary of the fit is shown below.

```
NO1 median <- NO1 %>%
  group by(nx*nfacet) %>%
  summarise(actual = median(value))
fit_lm2 <- lm(actual ~ poly(log(`nx * nfacet`) ,1, raw=TRUE), data = N01_median)</pre>
summary(fit lm2)
##
## lm(formula = actual ~ poly(log(`nx * nfacet`), 1, raw = TRUE),
##
       data = NO1_median)
##
## Residuals:
                      1Q
                             Median
                                             3Q
  -0.0021910 -0.0002941 0.0001063 0.0003966
##
## Coefficients:
##
                                             Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                            4.037e-02 3.814e-04 105.85
## poly(log(`nx * nfacet`), 1, raw = TRUE) 2.757e-03 8.043e-05
                                                                  34.27
                                                                            <2e-16
```

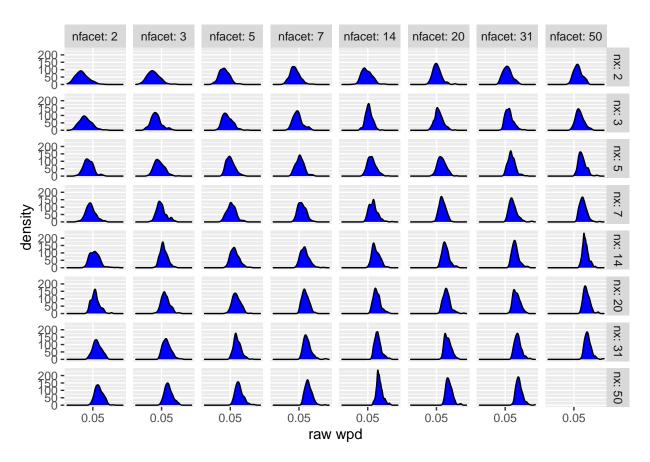


Figure 1: Distribution of raw wpd is plotted across different nx and nfacet categories. Both shape and scale of the distribution changes for different nx and nfacet categories.

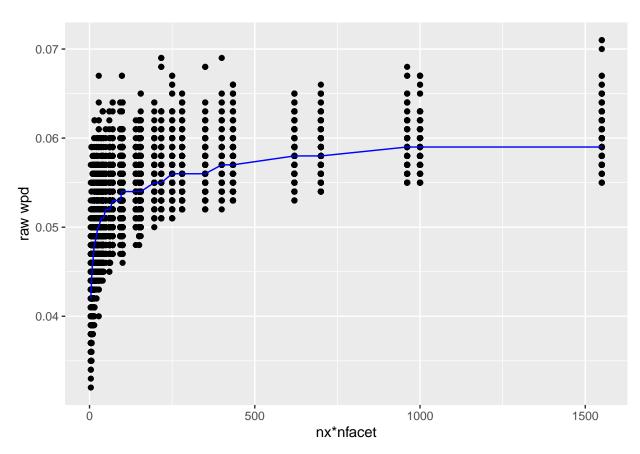
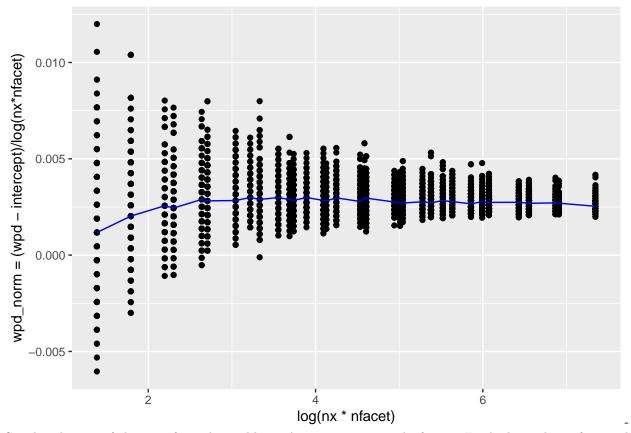


Figure 2: The raw wpd is plotted against nxnfacet and the blue line represents the median of the multiple values for each nxnfacet levels.

• Make it a horizontal line so that the values are not affected by nx*nfacet by plotting $wpd_{norm} = (value - intercept)/log(nx*nfacet)$ against log(nx*nfacet). The line actually becomes horizontal after this transformation as could be seen in Figure ??.



See distribution of the transformed variable wpd_{norm} across nx and nfacets. For higher values of nx and nfacet, the distribution is similar, but not for lower values.

It was suggested that we can do permutation normalization for smaller levels and this scalar transformation for higher levels as this transformation is working well when both nx and nfacet are high enough. There are potentially two problems if we follow that approach:

- Since the normalized wpd's for different harmonies will be on different scales, harmonies could not be ranked even for the same objects (households) to select the most important household for that object which was essentially the idea. Hence, the normalized wpd's would have to be brought on the same scale irrespective of the approach (whether permutation approach or scalar transformation).
- We have to think about the cut-offs, that is, for which level of nx and nfacets, we want to move to scalar transformation like this.

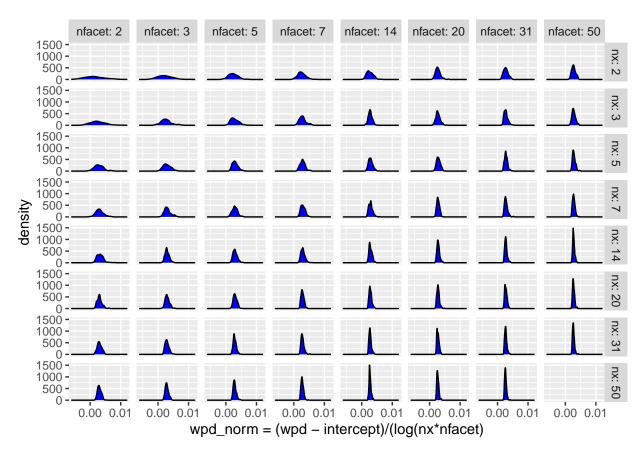


Figure 3: The distribution of the transformed wpd is plotted. The shape and the scale of the distribution still changes across different nx and nfacet levels, but seems uniform for higher levels.