

Data Quality Dashboard

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Agenda

Motivation

Quality indicators

Robust variance estimates

Outlier detection

Demo - Graphical User Interface



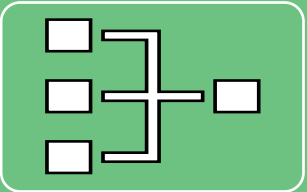
Motivation



Need for feedback on quality of newly established surveys of service industries



Develop a unified system for assessing data quality



New Python production system for currently SPPI being developed



Quality Indicators

Recommended Quality Indicators in Official Statistics (SSB 2024/5)

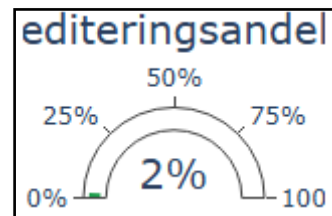
- Comparable to
 - Single Integrated Metadata Structure (ESS)
 - Generic Statistical Business Process Model (UN ECE)
- Chosen indicators for the dashboard:
 - Imputation rate
 - Data completeness
 - Unit response rate
 - Uncertainty (Coefficient of variation)
 - Control failure rate



Imputation rate

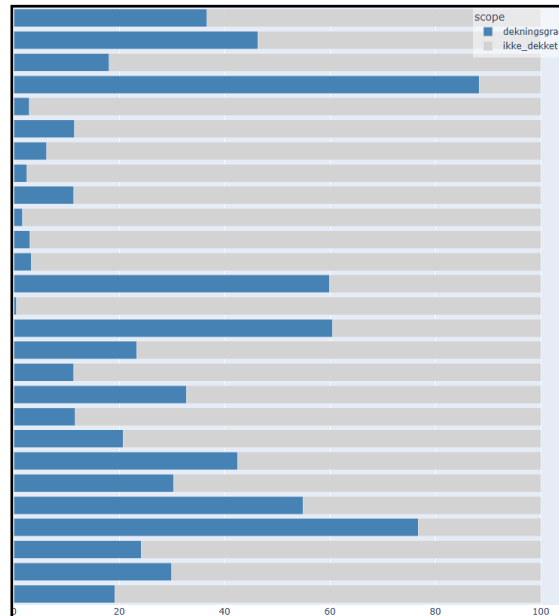
- Share of prices that has been changed. Either through manual correction or an automatic process.
- Requires that all individual data cells (price) must have associated process data about how it has been changed
- Interpretation : A high imputation rate might be an indicator that the accuracy of the received data is poor.

- Visual representation:



Data completeness

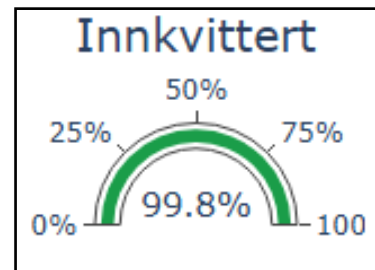
- Share of units with value (non missing) in relation to number of units in the data set.
- Target: Ideally we want the share of units with values to be as close to one as possible.
- Visual representation:



Unit response /Item response

- Share if units recieved in relation to the total units in data sett
- Target: Idealy we want the share to be as close to one as possible.

- Visual representation:



Uncertainty (Coefficient of variation)

Models for elementary indices

- Each Index can be expressed with a regression model
- Prices are random observations of an underlying inflation rate

- Carli: $I_{ij}^{s,t} = \frac{p_{ij}^t}{p_{ij}^s} = \theta_i + \epsilon_{ij}$

- Dutot: $p_{ij}^t = \theta_i p_{ij}^s + \epsilon_{ij}$

- Jevons: $\log I_{ij}^{s,t} = \mu_i + \epsilon_{ij}$



Variance

- Theoretical variance derived under assumptions like homoscedasticity (equal variances for all price relatives).
- Variance tells us how much individual prices disagree:
 - If all move together → variance = 0.
 - If prices move very differently → variance is larger.
- Example (Carli):

$$V(P) = \frac{\sigma_i^2}{n}$$



Plug-in Estimation

- Idea: Estimate variance from residuals and substitute into formula.
- Example (Carli):

$$\hat{\sigma}^2 = \frac{1}{n-1} \sum (I_{ij} - \hat{\theta}_i)^2$$

$$V(P) = \frac{\hat{\sigma}^2}{n}$$

- Works if all products share the same variance



Problem: Pricerelatives does not have same variance

- Theoretical formulas assume prices have the same variance.
- In reality, some products are more volatile than others.
- Robust method:
 - Look at the actual residuals (differences between observed and average).
 - Use them to estimate variance without strong assumptions.
- More realistic for messy real-world data?



Robust variance estimates (Zhang, 2010)

- Carli: $\hat{V}(P_i^{s,t}) = \frac{\sum_j e_{ij}^2}{n_i(n_i-1)}$
- Dutot: $\hat{V}(P_i^{s,t}) = (\sum_k p_{ik}^s)^{-2} \sum_j (1 - \frac{p_{ij}^s}{\sum_k p_{ik}^s})^{-1} e_{ij}^2$
- Jevons: $\hat{V}(P_i^{s,t}) = (P_i^{s,t})^2 \frac{\sum_j e_{ij}^2}{n_i(n_i-1)}$
- Variance estimates are now only dependent on the residual

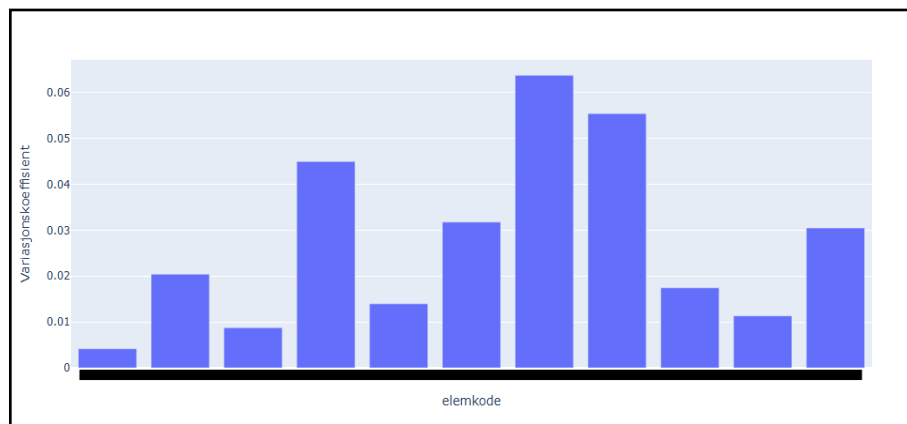


Coefficient of variation

- Relative measure of uncertainty
- Easier to compare variances between different service areas

$$CV = \frac{\sqrt{\text{var}(\hat{\theta})}}{\hat{\theta}}$$

- Visual representation:



Control failure rate

- Measurement: Share of units failing checks in relation to the number of units checked
- In this context, we consider outliers outside a threshold value to be a failure.
- Two ways used calculate outliers:
 - Studentized Residuals - R_{Student}
 - Difference in Fit Studentized – DIFFITS



Control failure rate continued

- Calculated using a simple linear regression model:

$$P_{i,t} = \beta P_{i,t-1} + \epsilon_i$$

$$\hat{\epsilon}_i = P_{i,t} - \beta P_{i,t-1}$$

- Studentized Residuals:

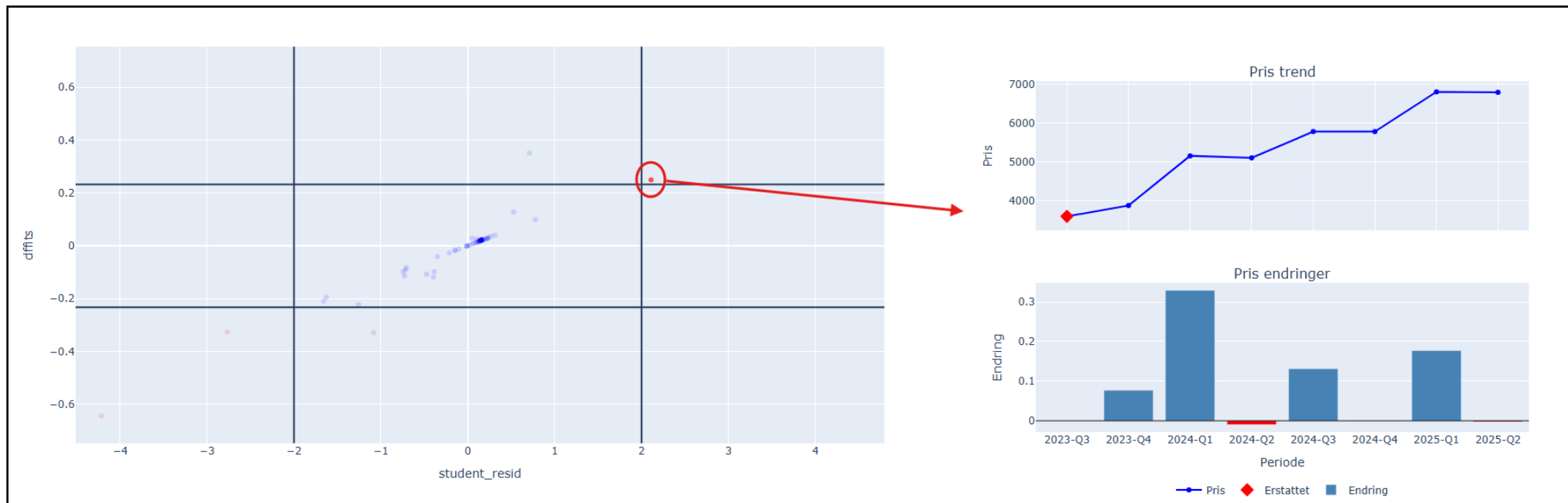
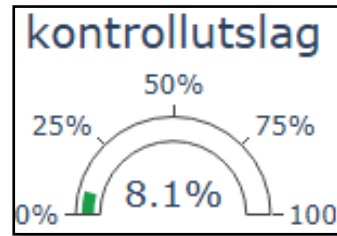
$$\frac{\hat{\epsilon}_i}{\sigma \sqrt{i - h_{ii}}} > 2$$

- Studentized Difference in fit:

$$\frac{\hat{y}_i - \hat{y}_{i(i)}}{s_i \sqrt{h_{ii}}} > 2 \sqrt{\frac{1}{n}}$$



Visual representation



Graphical User Interface

- Interactive dashboard for monitoring the quality indicators of our sample
- Background written using Python
 - Main programming language at Statistics Norway
 - Open source
 - Some SQL queries
- Dashboard made using the Dash framework
 - A package for making data-dashboards in Python
 - Some HTML and CSS



Data sources

- SQL data tables containing metadata on units in sample
- Price data mainly from surveys (other data sources are also represented)
- Production file – containing information on: type of imputation, base prices, codes for aggregation etc.



DEMO!



Thank you!

