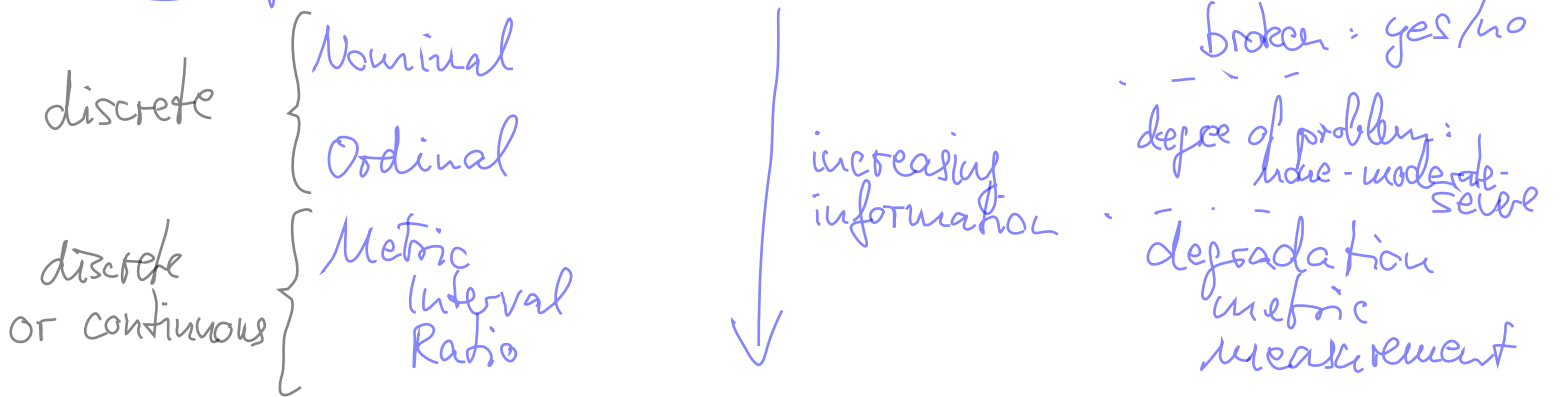


(SU 3 Nov 2021)

| | L1 | L2 | L3 | L4 | Sum |
|--------|------|------|------|-----|-----|
| young | | | | | 100 |
| medium | | | | | 100 |
| old | 35,5 | 46,2 | 14,0 | 4,3 | 100 |

35,5% of the oldest age group have education level 1.

Scales of measurement



Measures of central tendency

| Nominal | Ordinal | Metric |
|---|---------|--|
| Mode | Mode | Mode (sometimes difficult for continuous data) |
| problematic, if the distribution is very flat | | |
| A1 20 | A2 20 | A3 19 |

Median $x_{0,5}$
50%-quantile
middle value

Median $x_{0,5}$
50%-quantile

$$x_{0,5} = \begin{cases} x_{(\frac{n+1}{2})} \\ \frac{x_{(\frac{n}{2})} + x_{(\frac{n}{2}+1)}}{2} \end{cases}$$

n odd

n even

The median is robust

$x_{(1)}$ = smallest data value ... $x_{(n)}$ largest data value

Measures of central tendency only for metric data:

Arithmetic mean: $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$ unrobust

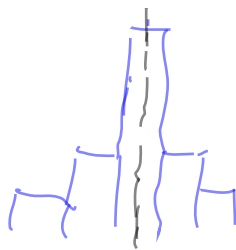
Robustified version: trimmed mean

omit a percentage of values at both ends of the scale

Relation between mode, median and mean can be used for assessing shape of a distribution:



left-skewed
mode > median > mean



symmetric
mode = median = mean



right-skewed
mode < median < mean

Measures of variability (or "dispersion")

only for metric variables

Range: $x_{(n)} - x_{(1)}$ (largest - smallest value)
remember: parentheses in index denote ordered values
very unrobust

For the next metric, we need to define quantiles: The **p-quantile** x_p is a value that separates the smallest p from the largest 1-p
(p a proportion between 0 and 1)
0% and 100%

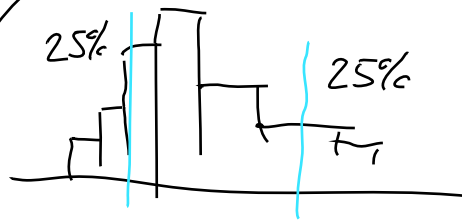
We already saw $x_{0.5} = x_{50\%}$, the median.

One (of many) formula for x_p :

$$x_p = \begin{cases} x_{(k)} & n \cdot p \text{ not an integer,} \\ & k = \lceil n \cdot p \rceil \text{ (ceil}(n \cdot p)) \\ \frac{x_{(k)} + x_{(k+1)}}{2} & n \cdot p = k \text{ an integer} \end{cases}$$

$$IQR = x_{0.75} - x_{0.25}$$

width of the middle 50% of the data



robust

upper and lower quartile

Mean absolute deviation from mean

$$\frac{1}{n} \sum_{i=1}^n |x_i - \bar{x}|$$

Mean absolute deviation from median

$$\frac{1}{n} \sum_{i=1}^n |x_i - x_{0.5}|$$

Median minimizes $\sum_{i=1}^n |x_i - m|$

w. r. t. m

Shortest average distance from a center value that is achievable with these data

Median absolute deviation from median

...

Empirical variance:

$$\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

Here, \bar{x} minimizes $\sum_{i=1}^n (x_i - m)^2$ w.r.t. m

This is one reason
that \bar{x} is so unrobust:
square emphasizes
large deviations

Variance

- is very important
- is very unrobust
- is in squared units of the data
- hard to interpret
- moment of inertia

Standard deviation

- square root of variance
- \Rightarrow is again in units of the data