



The European Commission's science and knowledge service

Joint Research Centre

Step 4: Normalisation for Composite Indicators

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Ten Steps Guide for Composite Indicators & Scoreboards

**Next
version**



OECD



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➤ *Next version (2019-2020)*

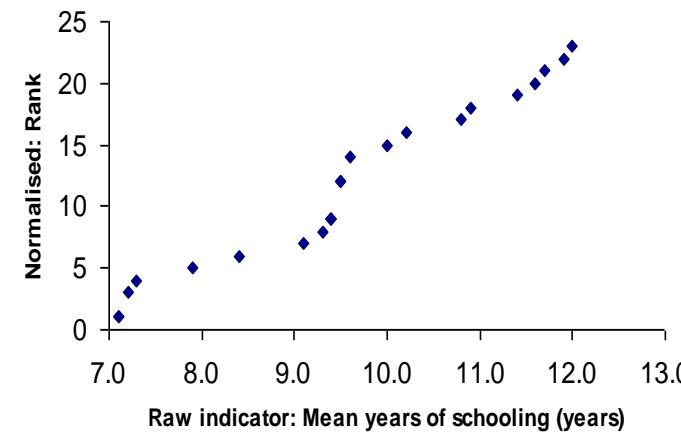
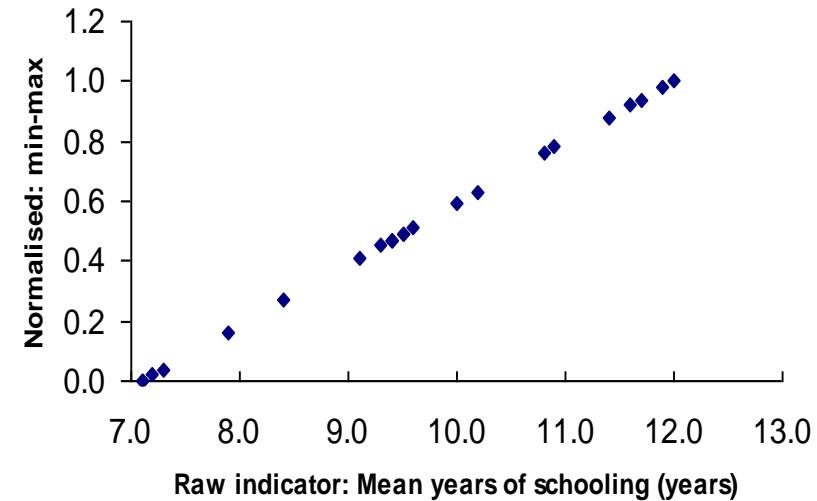
- Step 10. Visualisation & Communication
- Step 9. Back to the data
- Step 8. Robustness & Sensitivity
- Step 7. Statistical coherence
- Step 6. Aggregation
- Step 5. Weighting
- Step 4. Normalisation
- Step 3. Data treatment
- Step 2. Selection of indicators
- Step 1. Developing the framework



Step 4

Normalisation for C.I.

- ✓ Make **directional adjustments**
(so that higher scores correspond to better performance in all indicators or vice versa)
- ✓ Select a **suitable normalisation method** that respects the conceptual framework and the data properties



Outline

- What is normalisation?
- Prior to normalisation
- Normalisation methods: min-max, distance to a reference, standardisation (or z-score), categorical scale, ranking and quantile empirical distribution
- Summary table
- Key messages

What is normalisation?



Adjustments of distribution and scale of variables

It allows for aggregation of variables by averages (i.e. composite indicators) and for comparisons and robust aggregations

- different units of measurement => **common scale**
- different ranges of variation => **similar range of variation**

Different normalisation methods lead to different results
Choice based on the general objective

Normalising by...

Sample statistics

*To obtain comparability across variables
adjusting for:*

1. different nature of indicators (positive vs. negative orientation towards the index)
2. different units of measurement across indicators
3. different ranges of variation

A common variable

To obtain comparability across countries

Dividing the raw data by size of population, land area, gross domestic product (GDP), or other **denominator** to make data comparable across countries

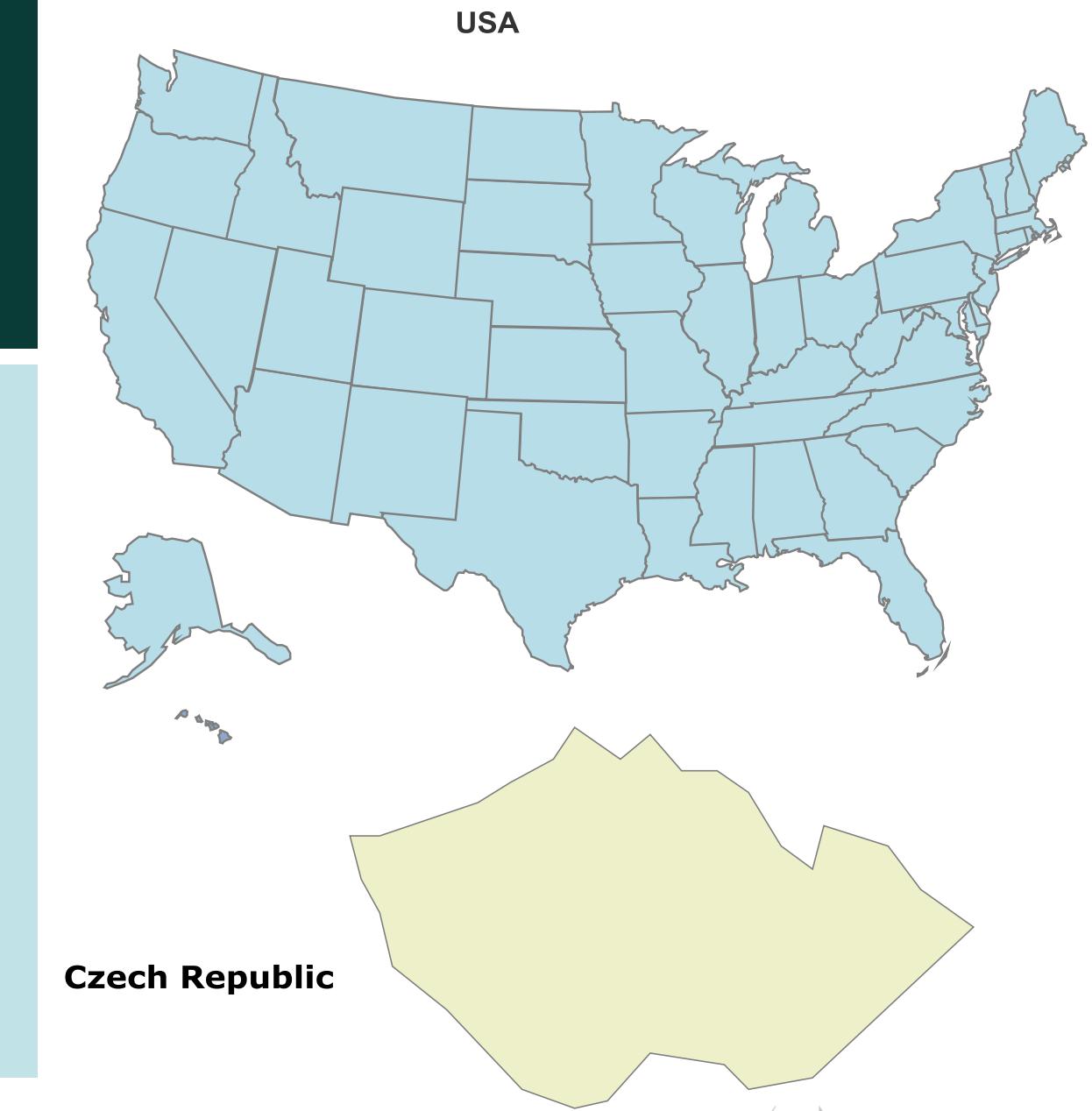
Prior to normalisation

Comparisons across countries

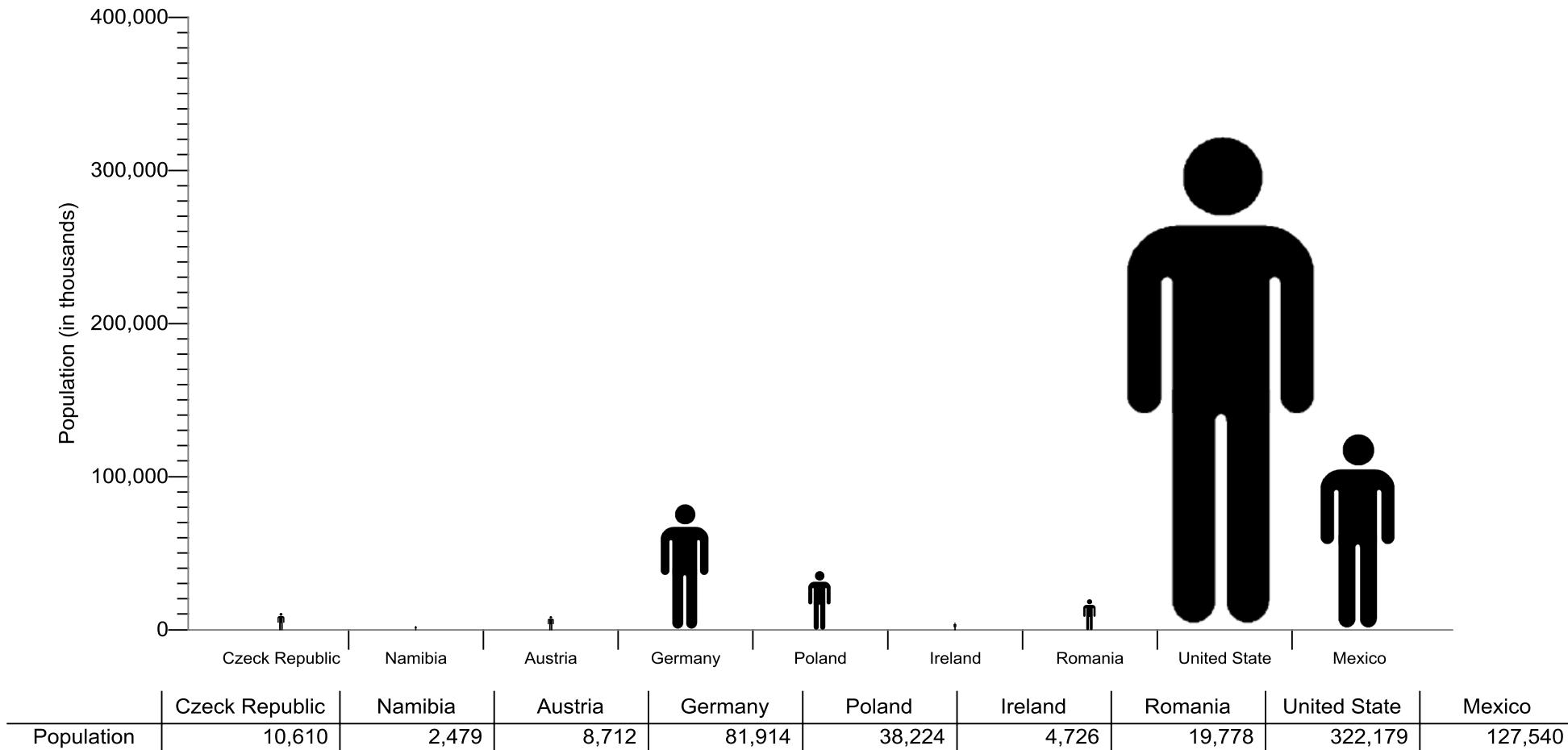
Q: What to consider?

US, 322 Million inhabitants

Czech Republic, 10 Million inhabitants



World Population in 2016 (thousand)



Source: United Nations, 2016

Beer consumption - liters per capita	Czech Republic	Namibia	Austria	Germany	Poland	Ireland	Romania	US	Mexico
2016	143.3	108.0	106.0	104.0	100.8	98.2	94.1	74.8	65.1

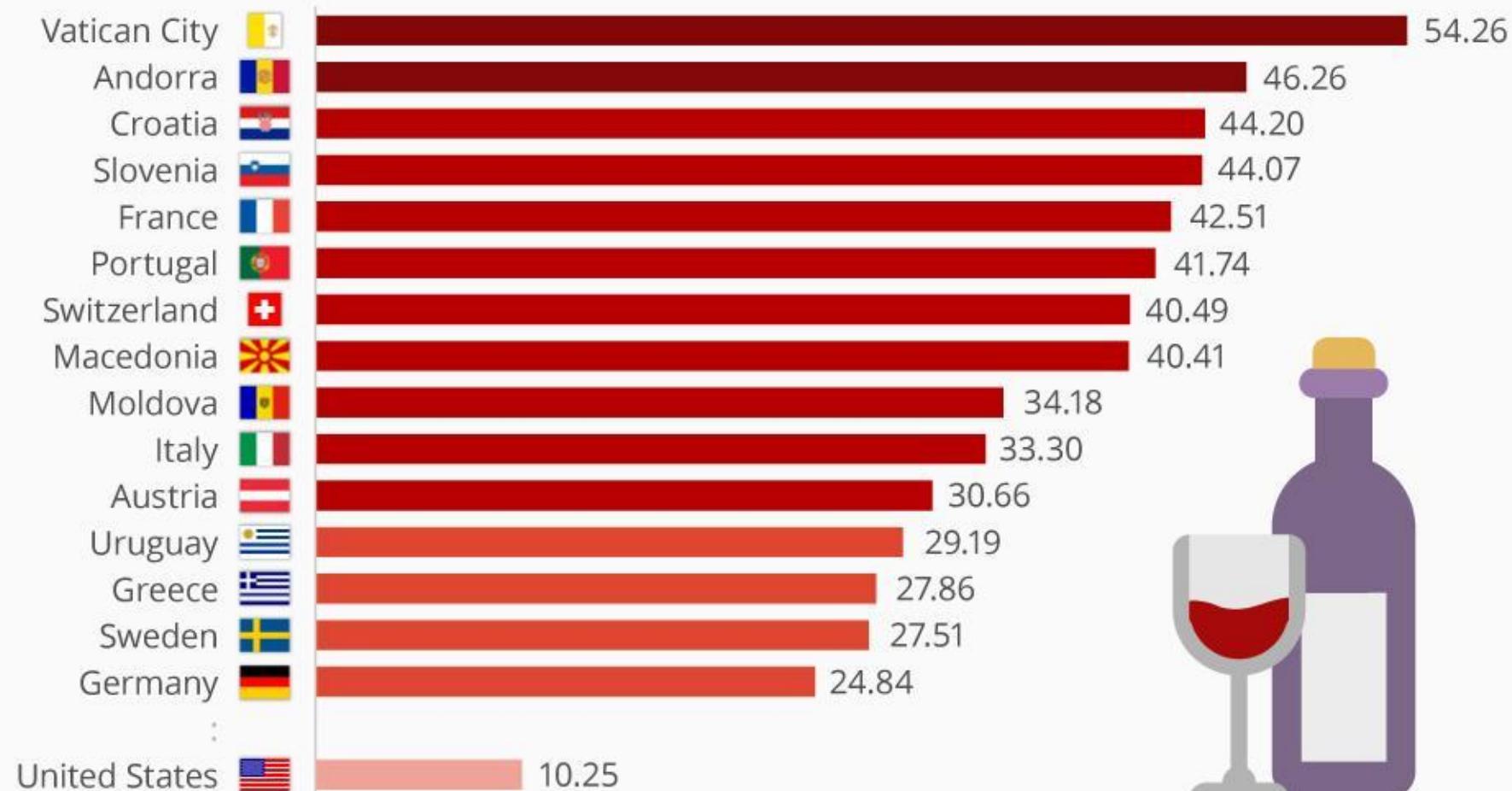
World Beer Consumption per capita



https://www.kirinholdings.co.jp/english/news/2017/1221_01.html

The World's Biggest Wine Drinkers

Annual per capita wine consumption worldwide (Nov 15)*



* Liters per capita, excluding overseas territories

Source: The Wine Institute

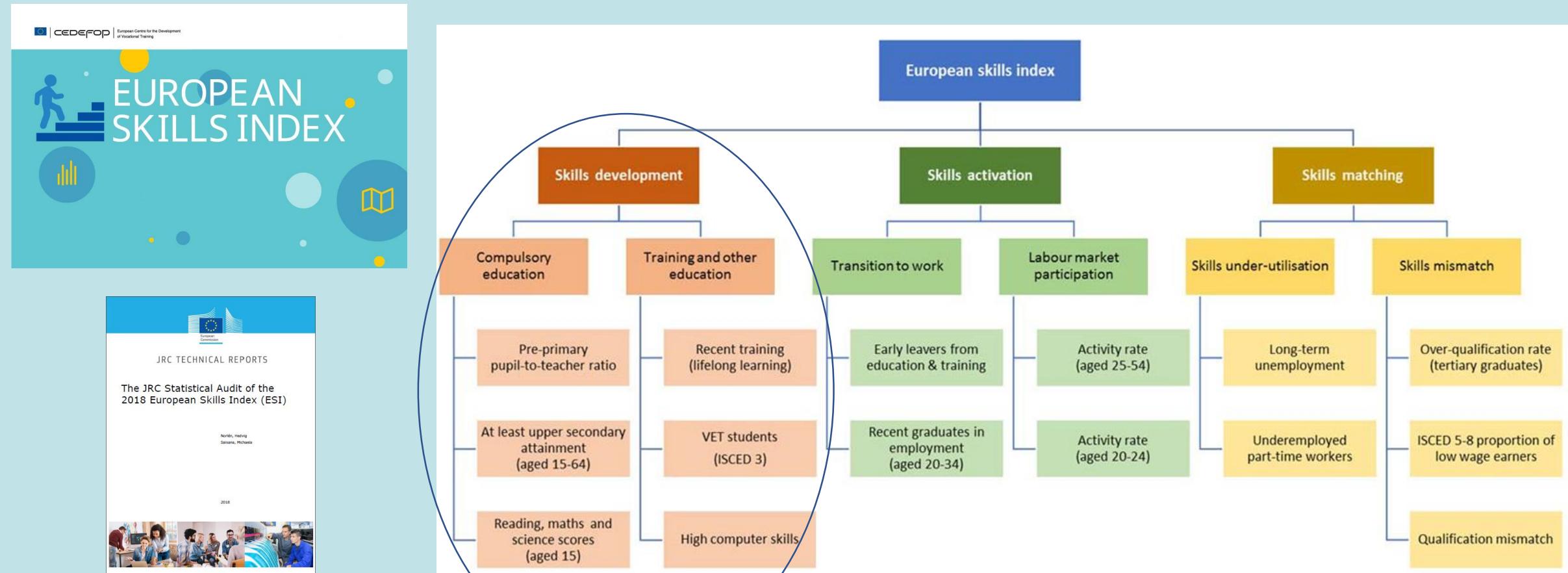
Forbes statista

Data normalisation



- It is about changing the scale of the data
- Common practice: scaling data between $[0, 1]$ or $[0, 100]$
- Mixed scales (e.g., European skills Index: PISA scores and unemployment rate have different ranges)

Example: European Skills Index



Range of variation

Example: European Skills Index



Table 4.2: Summary statistics

Indicator (unit)	Range
Pre-primary pupil-to-teacher ratio (students per teacher)	[6.4, 21.5]
Share of population aged 15-64 with at least upper secondary education (%)	[47.1, 87.6]
Reading, maths and science scores (aged 15) (PISA score)	[437.7, 524.3]
Recent training (%)	[1.2, 29.6]
VET students (%)	[1.2, 73.2]
High computer skills (%)	[7.0, 46.0]

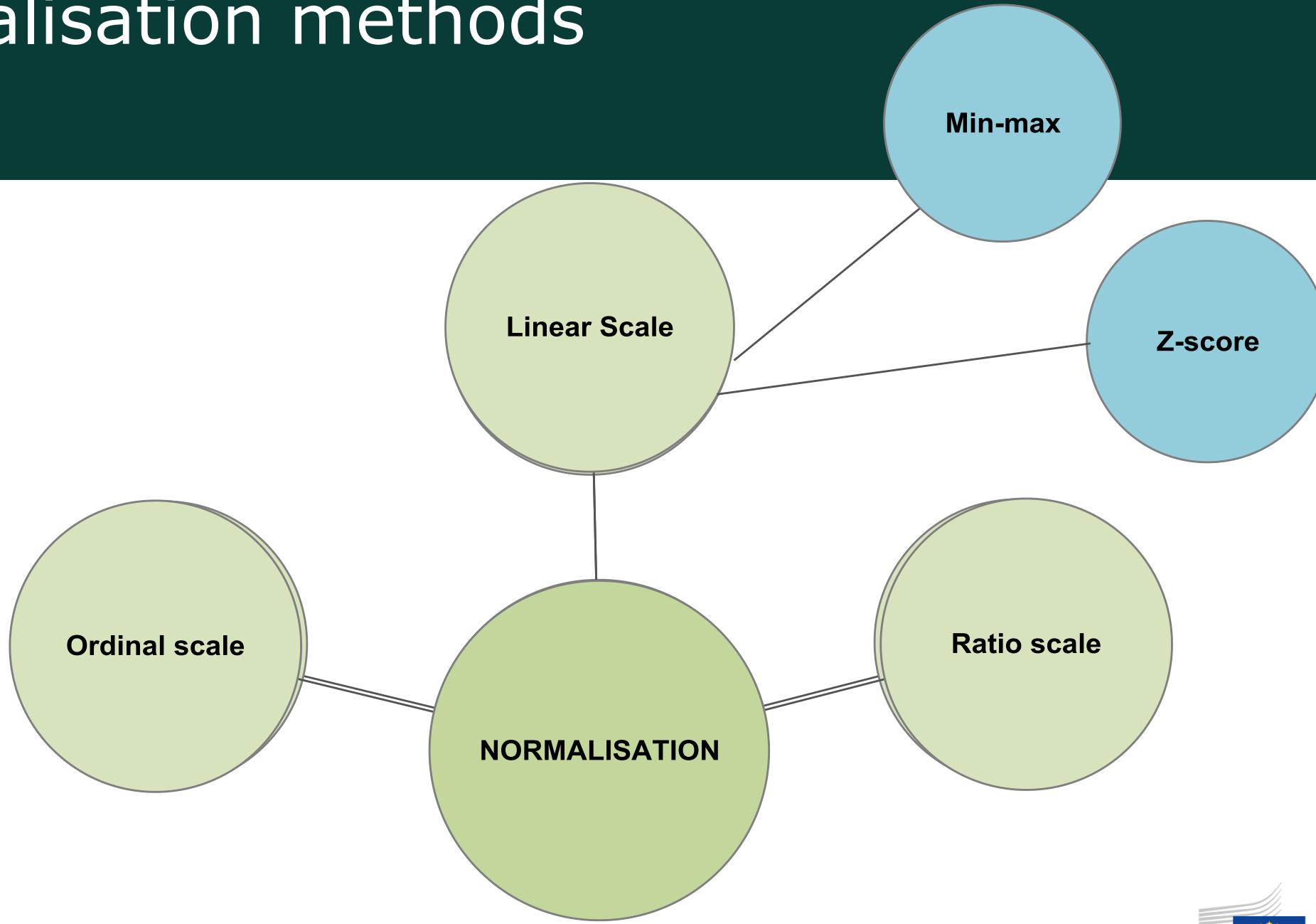


W. Edwards Deming

If you change the rule for counting people, you come up with a new number.

source: quotes.deming.org/10217

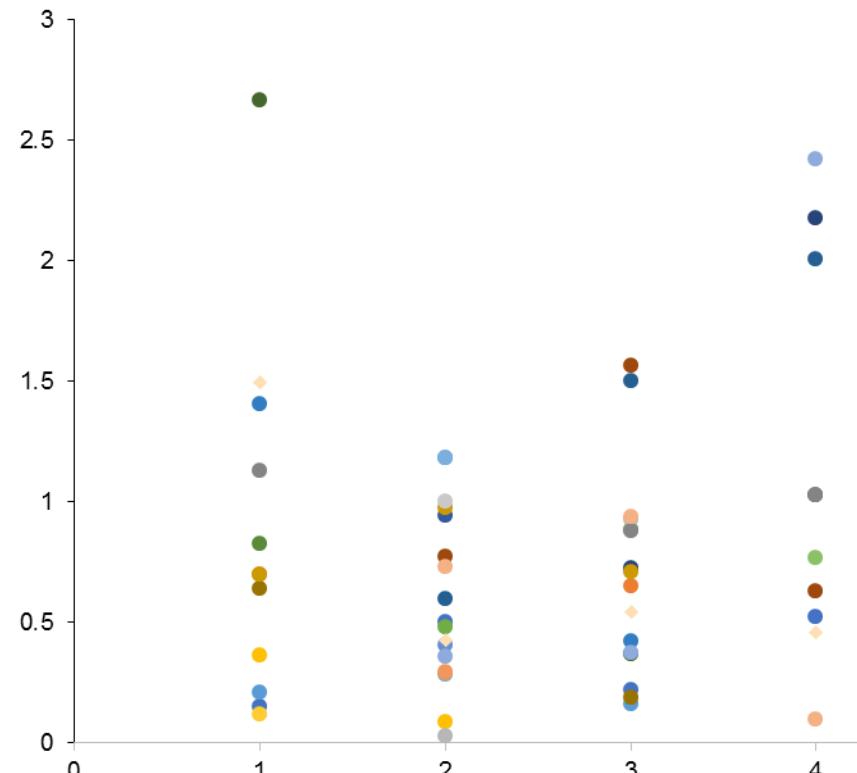
Normalisation methods



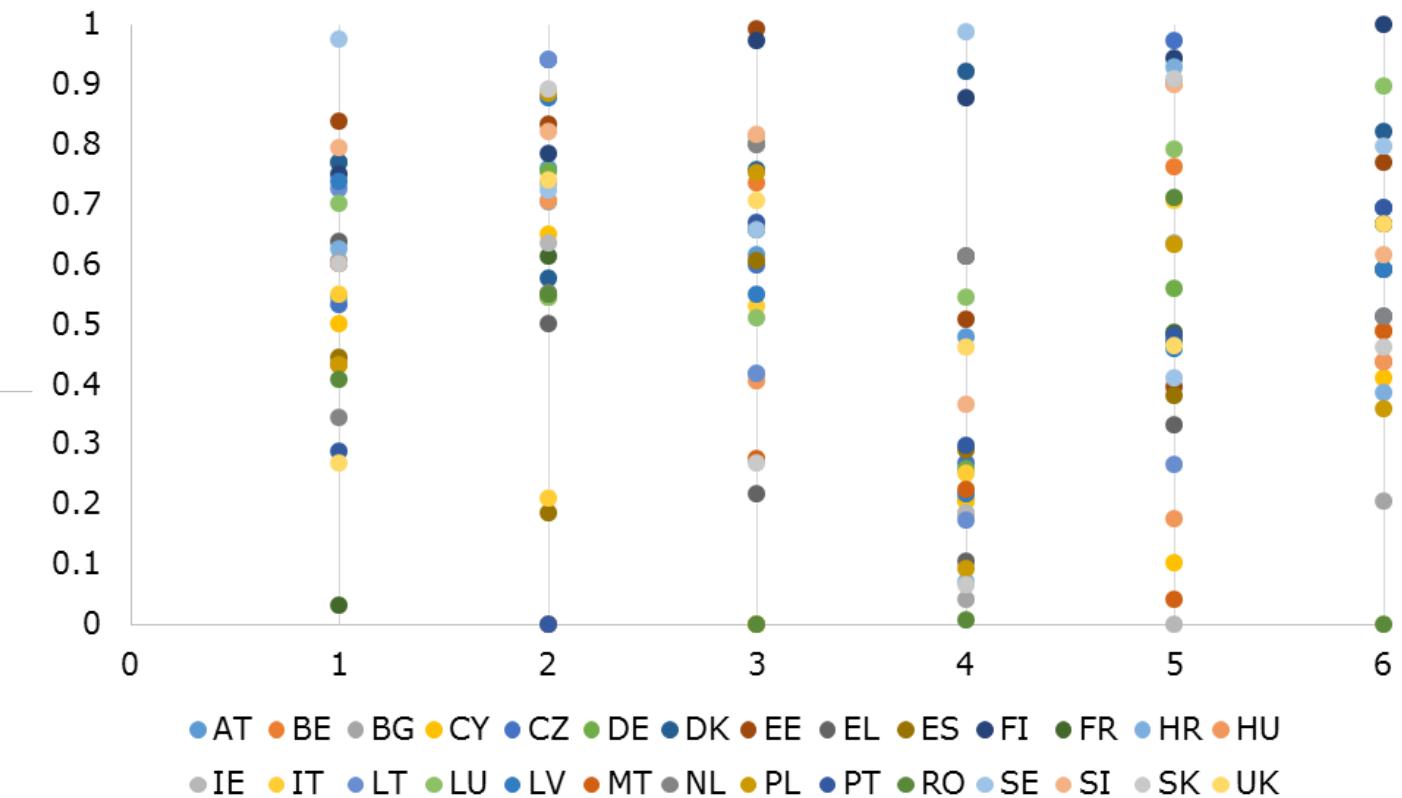
Data before normalization



Z-score



Min-max

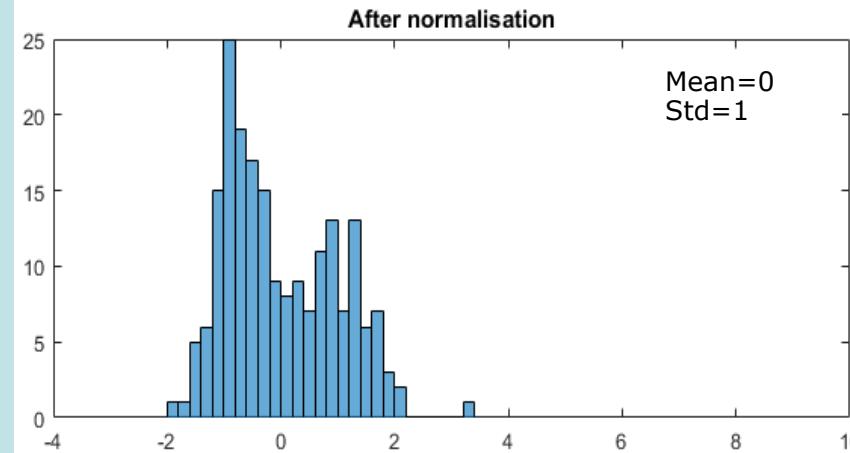
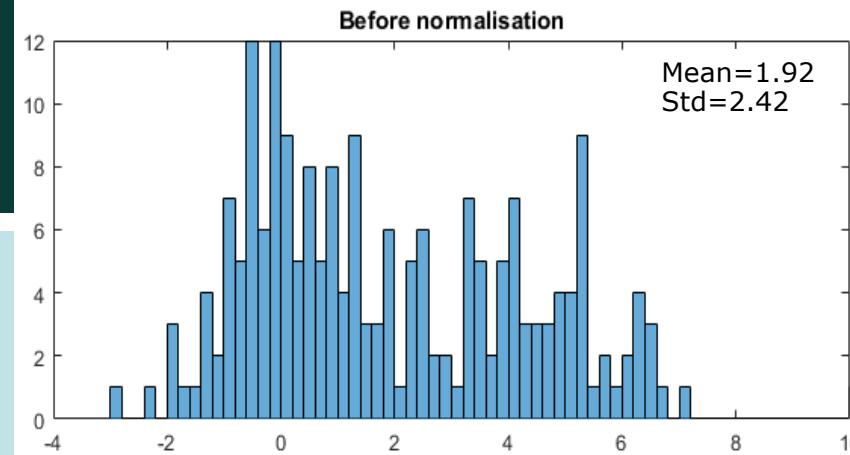


Normalisation methods – linear scale Standardisation (Z-score)

$$I = \frac{x - \bar{x}}{S} \quad \sim \quad z = \frac{x - \mu}{\sigma}$$

- Imposes a distribution with mean zero and variance 1
- **Normalised indicators have:**
 - ✓ same variance (=1)
 - ✓ not necessarily same range of variation
- **Method:** sensitive to extreme values/outliers

Standardised scores which are below average are negative => implications on the use of geometric average as an aggregation method



Normalisation methods – linear scale, Min-max

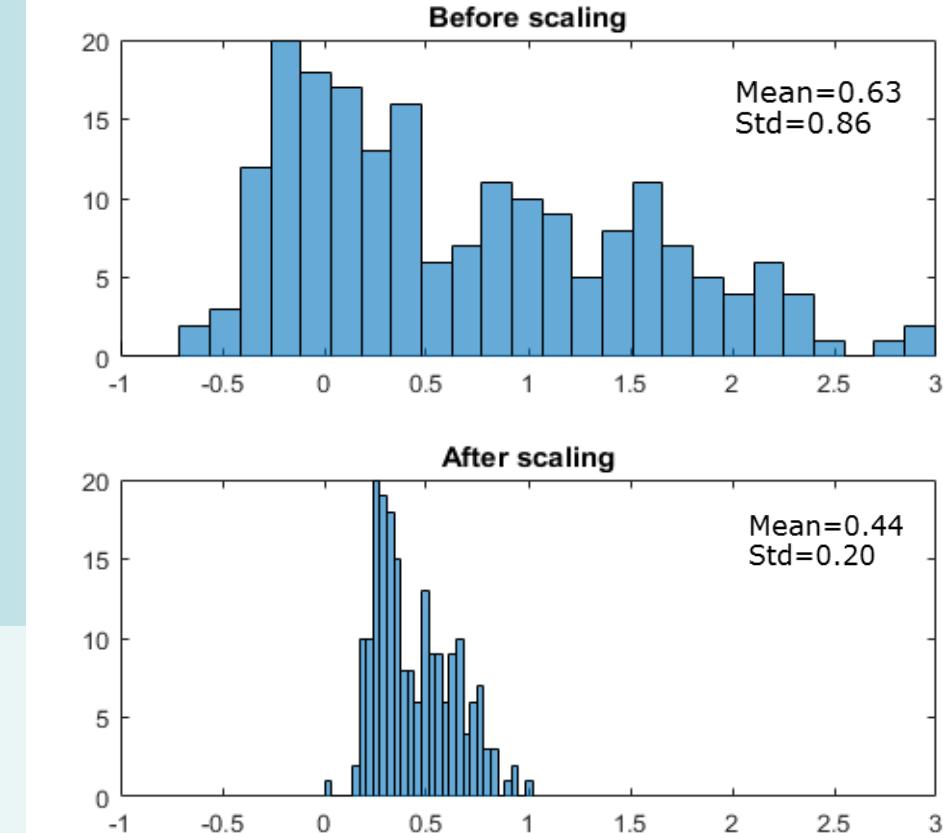
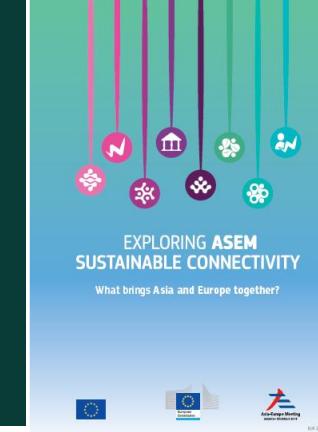
$$I = \frac{x - \min}{\max(x) - \min(x)}$$

Normalised indicators have:

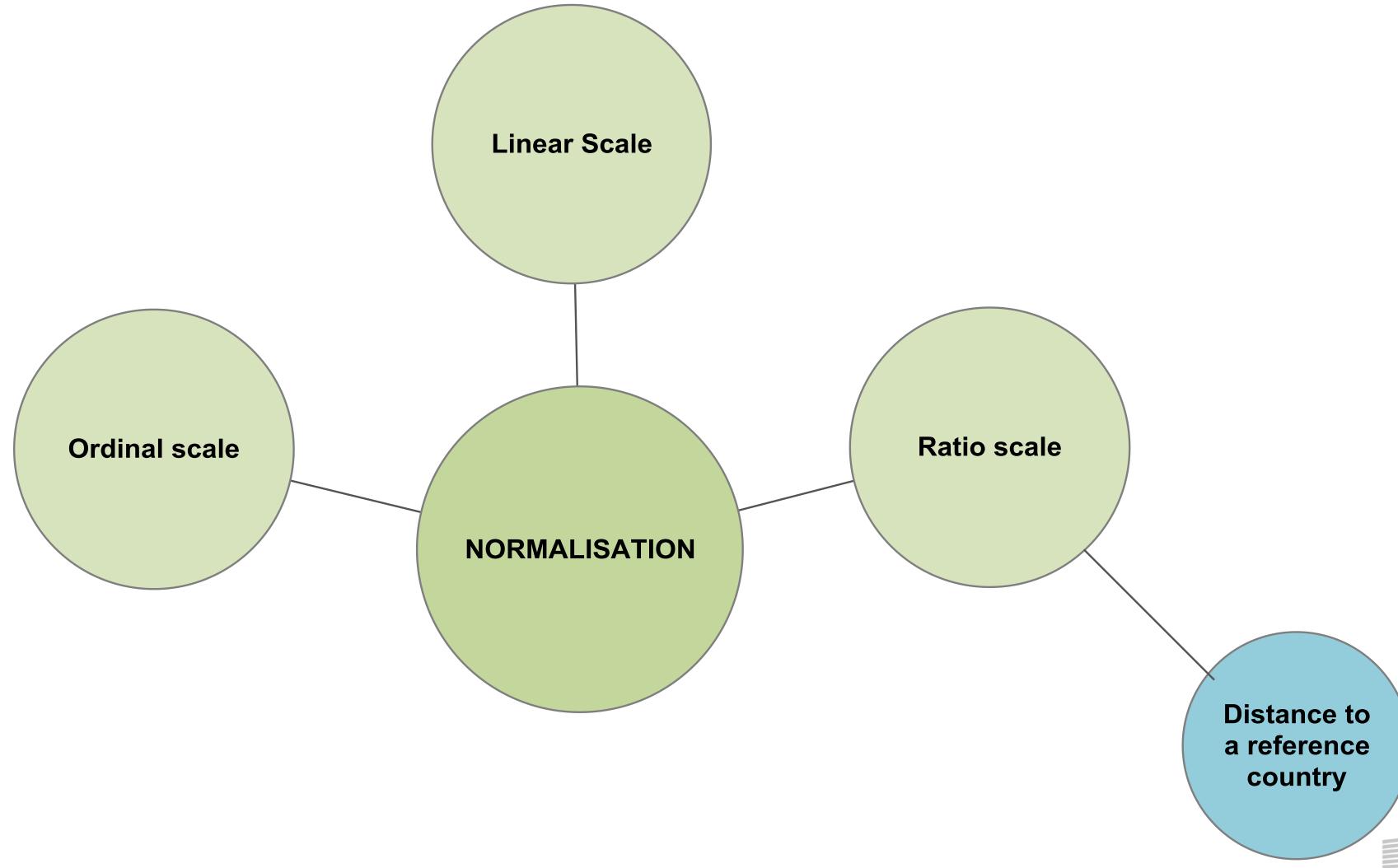
- ✓ same range of variation e.g., [0,1]
- ✓ not same variance

Method: sensitive to extreme values/outliers

Rescaling is easier to communicate to a wider public because it normalises indicators to an identical range [0, 1], [0, 100], where usually higher scores represent better achievement.



Normalisation methods: Ratio scale



Normalisation methods: Ratio scale - Distance to a reference country

The reference country can be:

- the group leader or an external benchmark country
- a hypothetical country (target to be reached in a given timeframe)
- or an aggregate/average (eg., EU28, world)

The reference value could be fixed at a specific point in time (e.g. initial t_0) to account for the indicator evolution across time

Normalised indicators do not have:

- ✓ same range
- ✓ same variance

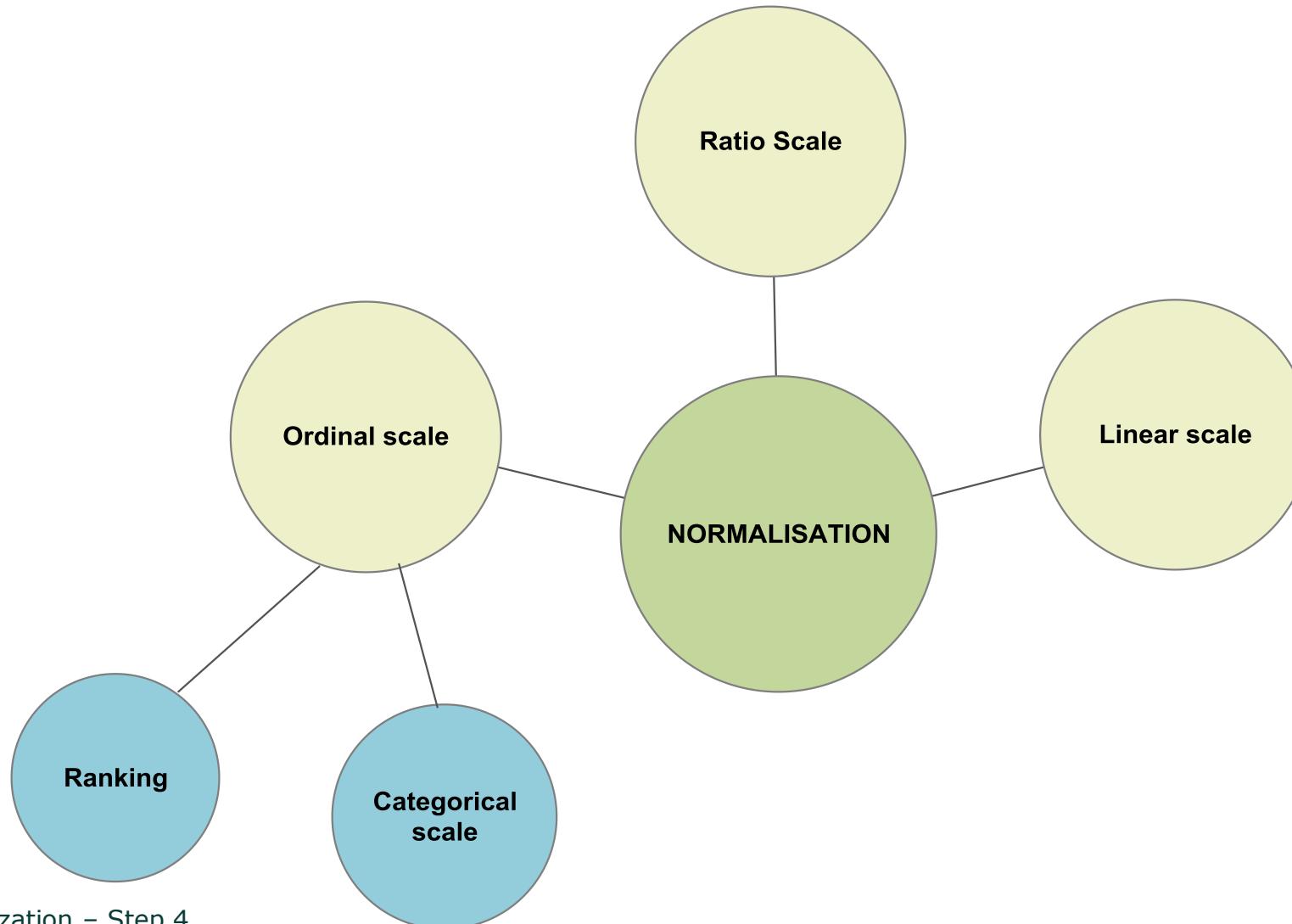
Method: sensitive to extreme values/outliers

$$I_c = \frac{x_c}{\bar{x}_c}$$

$$I_c = \frac{x_c^t}{x_c^{t_0}}$$



Ordinal scale: Ranking across countries and categorical scale

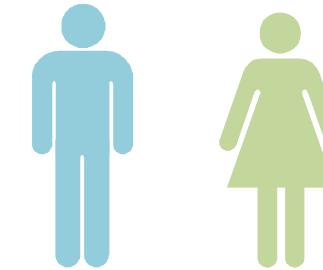


Normalisation methods – Ordinal scale

Categorical scales

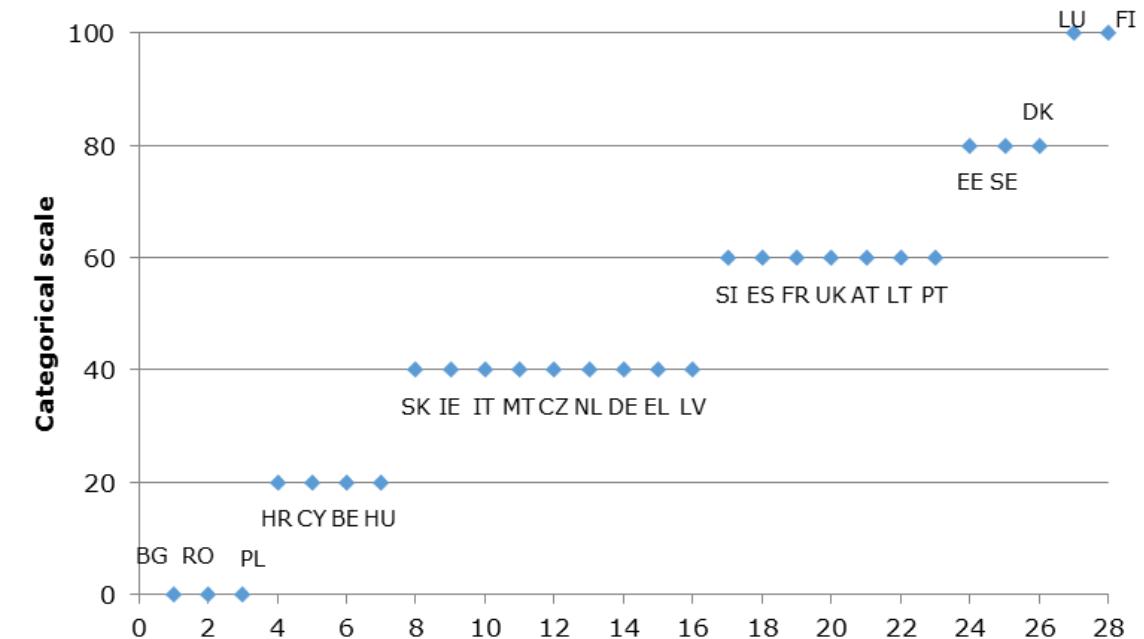
- The score of each indicator is based on categories (nominal or numerical)
- Each category represents a portion of the range of the variable
- Each category score can be based on the percentile of the distribution of the indicator across countries
- Remember to justify the choice of intervals and scores
- **Normalised indicators have:**
 - Same range [0, 100]
 - Same variance: if there are NOT tied ranks
- **Method:** not sensitive to extreme values/outliers
- **Distribution:** NO uniform

Nominal scales (e.g. gender, blood type: A, B, AB, O)



Numerical scales

$$I_{q,c}^t = \begin{cases} 0 & \text{if } p^0 \leq x_{q,c}^t < p^{10} \\ 20 & \text{if } p^{10} \leq x_{q,c}^t < p^{25} \\ 40 & \text{if } p^{25} \leq x_{q,c}^t < p^{50} \\ 60 & \text{if } p^{50} \leq x_{q,c}^t < p^{75} \\ 80 & \text{if } p^{75} \leq x_{q,c}^t < p^{90} \\ 100 & \text{if } p^{90} \leq x_{q,c}^t \leq p^{100} \end{cases}$$



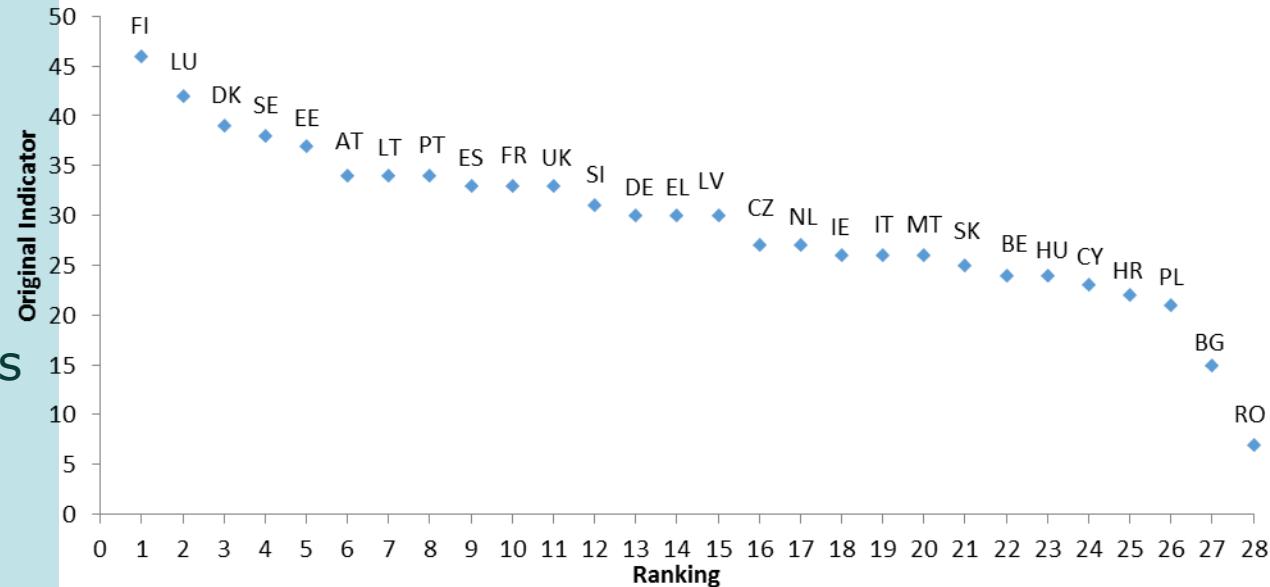
Normalisation methods – Ordinal scale Ranking across countries

- Scores are replaced by ranks – e.g. the highest score receives the first ranking position (rank 1)
- Uses ordinal information only – information on levels is not maintained
- **Normalised indicators have:**
 - ✓ the same range: $[1, n]$ ($n = \text{no. of countries}$)
 - ✓ the same variance 67.67 ($n=28$)
- **Method:** not sensitive to extreme values/outliers
- Distribution: Uniform

$$I = \text{rank}(x)$$



High computer skills Ranking



Source: European Skills Index, 2018

Normalisation methods: Quantile empirical distribution

Definition: quantile normalization is a technique for making two distributions identical in statistical properties

$$u = \frac{\text{rank}(x)}{N + 1}$$

- where N is the sample size;
- $\text{rank}(x)$ is the rank associated to each realization

The quantile normalisation method allow to gather a variable range [0, 1]

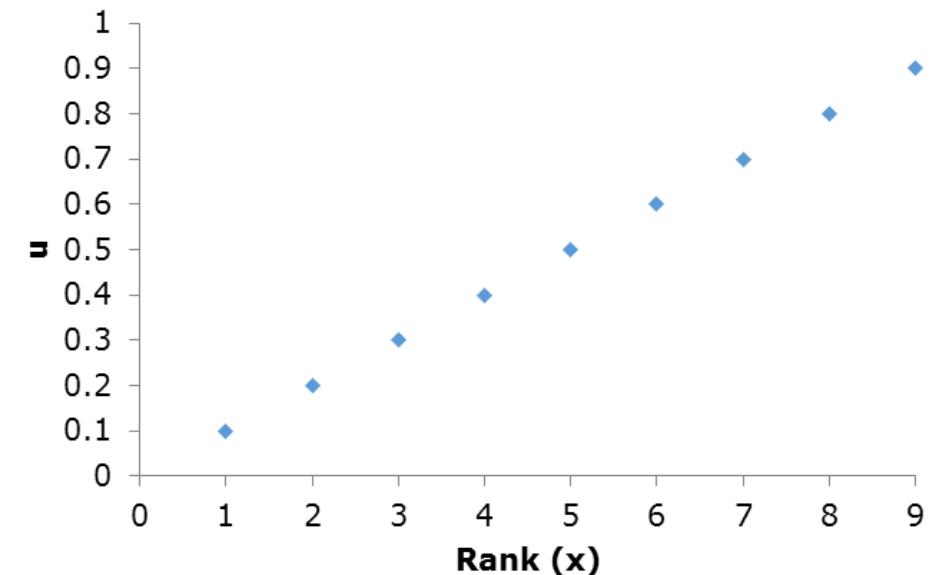
Normalisation methods: Quantile empirical distribution

For example: if $x = (10, 2, 8, 81, 38, 4, 19, 322, 127)$

Then, rank $(4, 1, 3, 7, 6, 2, 5, 9, 8)$; $u = (0.4, 0.1, 0.3, 0.7, 0.6, 0.2, 0.5, 0.9, 0.8)$ giving to 322 the 9th rank

$$u = \frac{\text{rank}(x)}{N + 1}$$

Example: let's consider the population in million by country – slide 9



$x = \text{million inhabitants}$	Rank(x)	u
2	1	0.1
4	2	0.2
8	3	0.3
10	4	0.4
19	5	0.5
38	6	0.6
81	7	0.7
127	8	0.8
322	9	0.9

Summary table

Adjustments/Method	Quantile empirical distribution	Ranking	Categorical scale	Z-score	Min-max	Distance to a reference country
Unit of measurement	Y	Y	Y	Y	Y	Y
Variance	Y	Y	Y/N	Y	N	N
Range of variation	Y	Y	Y	N	Y	N
Extreme values*	Y	Y	Y	N	N	N
Distribution*	Y	Y	Y/N	N	N	N
Extreme values*	non-sensitive to extreme values					
Distribution*	the distribution will be the same for the normalised indicators					

Yes is valid only if there are not tied ranks

Key messages

- There are different normalisation methods leading to different results
- Their selection depends on the data structure and what we want to achieve: e.g., **same variance=> Z-score; same range of variation =>min-max.** It is crucial to understand the properties of each
- Alternative normalisation methods can be included as an additional dimension in the uncertainty/sensitivity analysis
- Normalisation is the step the lowest impact on the ranking (based on the majority of the case studies)

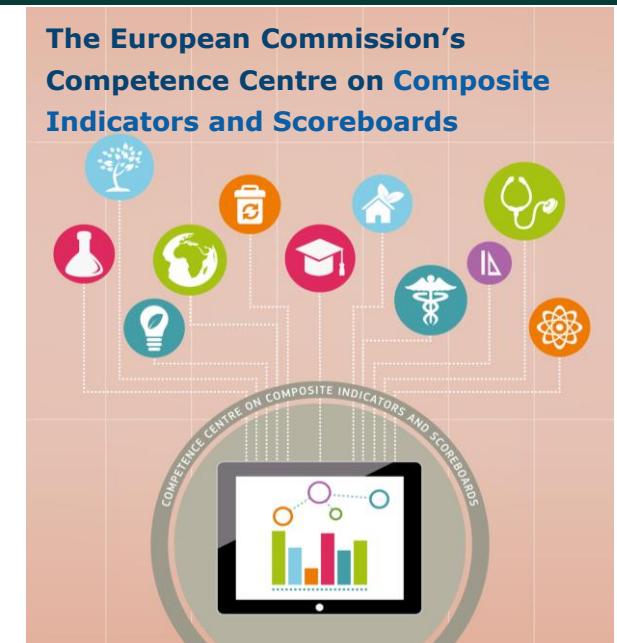


THANK YOU

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COIN tools are available at:
<https://composite-indicators.jrc.ec.europa.eu/>



References

- Becker, W., M. Saisana, P. Paruolo, and I. Vandecasteele. 2017. 'Weights and Importance in Composite Indicators: Closing the Gap. *Ecological Indicators* 80: 12–22.
- Bichindaritz I., Big Data, Genes, and Medicine, the State University of New York (SUNY), <https://www.coursera.org/lecture/data-genes-medicine/data-normalization-jGN7k>
- Cohen, A., Saisana, M., 2014, Quantifying the Qualitative: Eliciting Expert Input to Develop the Multidimensional Poverty Assessment Tool, *Journal of Development Studies* 50(1): 35-50.
- European Union, European Centre for the Development of Vocational Training (Cedefop). 2018, European Skills Index technical report. Thessaloniki, Greece.
- Forbes-Statista, 2016, The world's biggest wine drinkers, <https://www.forbes.com/sites/niallmccarthy/2016/05/17/the-worlds-biggest-wine-drinkers-infographic/#2c1468a77a57>
- Kirin Holdings Company. Kirin Beer University Report Global Beer Consumption by Country in 2016, Global Beer Consumption by Country in 2016, Table 1 [Per Capita Beer Consumption by Country in 2016](#), Table 3, https://www.kirinholdings.co.jp/english/news/2017/1221_01.html
- OECD/JRC, 2008, *Handbook on Constructing Composite Indicators. Methodology and user Guide*, OECD Publishing, ISBN 978-92-64-04345-9.

References

- Organisation Internationale de la Vigne ed du Vin, 2016, OIV, State of the viticulture World Market, p. 10.
- Paruolo P., Saisana M., Saltelli A., 2013, Ratings and Rankings: voodoo or science?. J Royal Statistical Society A 176(3), 609-634.
- Saisana, M., and Saltelli, A., 2011, Rankings and Ratings: Instructions for use, Hague Journal on the Rule of Law 3(2), 247-268.
- Saisana M., D'Hombres B., Saltelli A., 2011, Rickety Numbers: Volatility of university rankings and policy implications. Research Policy 40, 165-177.

Technical Appendix



Global Beer Consumption by Country in 2016

2016 Ranking	2015 Ranking	Country	2016			2015	
			Total Consumption (thousand kl)	Global Market Share	Growth Rate 2015-2016	Total Consumption (thousand kl)	Global Market Share
1	1	China	41,772	22.4%	-3.4%	43,264	23.0%
2	2	United States	24,245	13.0%	0.6%	24,106	12.8%
3	3	Brazil	12,654	6.8%	-2.7%	13,008	6.9%
4	5	Germany	8,412	4.5%	-0.5%	8,450	4.5%
5	4	Russia	8,405	4.5%	-1.8%	8,559	4.6%
6	6	Mexico	7,988	4.3%	8.4%	7,371	3.9%
7	7	Japan	5,251	2.8%	-2.4%	5,380	2.9%
8	8	United Kingdom	4,373	2.3%	-0.9%	4,413	2.3%
9	9	Vietnam	4,117	2.2%	7.4%	3,832	2.0%
10	11	Spain	3,909	2.1%	2.3%	3,821	2.0%
11	10	Poland	3,892	2.1%	1.8%	3,823	2.0%
12	12	South Africa	3,145	1.7%	2.4%	3,072	1.6%
13	13	India	2,701	1.4%	9.9%	2,457	1.3%
14	14	Colombia	2,357	1.3%	3.0%	2,289	1.2%
15	15	South Korea	2,160	1.2%	1.0%	2,139	1.1%
16	17	Canada	2,093	1.1%	-0.3%	2,100	1.1%
17	18	France	2,061	1.1%	2.4%	2,012	1.1%
18	19	Czech Republic	1,959	1.0%	1.5%	1,930	1.0%
19	20	Thailand	1,910	1.0%	1.5%	1,881	1.0%
20	23	Romania	1,826	1.0%	1.6%	1,797	1.0%
21	22	Argentina	1,778	1.0%	-5.2%	1,875	1.0%
22	25	Italy	1,745	0.9%	0.8%	1,730	0.9%
23	21	Ukraine	1,743	0.9%	-7.2%	1,878	1.0%
24	24	Australia	1,735	0.9%	0.1%	1,735	0.9%
25	27	Philippines	1,620	0.9%	6.0%	1,528	0.8%

Source: Kirin Company, Limited, (2016)

Global Wine Consumption

Main wine-consuming countries^a

mhl	2012	2013	2014	2015 ^b	2016 ^c	2016/2015	2016/2015
						Variation in volume	Variation in %
United States	30,0	30,2	30,4	31,0	31,8	0,8	2,5%
France	28,0	27,8	27,5	27,2	27,0	-0,2	-0,7%
Italy	21,6	20,8	19,5	21,4	22,5	1,1	5,3%
Germany	20,3	20,4	20,2	20,6	20,2	-0,4	-1,8%
China	17,1	16,5	15,5	16,2	17,3	1,1	6,9%
United Kingdom	12,8	12,7	12,6	12,7	12,9	0,2	1,4%
Spain	9,9	9,8	9,9	10,0	9,9	0,0	-0,4%
Argentina	10,1	10,4	9,9	10,3	9,4	-0,9	-8,3%
Russia	11,3	10,4	9,6	9,3	9,3	0,0	0,3%
Australia	5,4	5,4	5,4	5,3	5,4	0,1	2,4%
Canada	4,9	4,9	4,7	4,9	5,0	0,0	3,1%
Portugal	5,0	4,8	4,7	4,8	4,8	0,1	0,1%
South Africa	3,6	3,7	4,0	4,2	4,4		
Romania	4,3	4,6	4,7	3,9	3,8	-0,2	-4,5%
Japan	3,1	3,4	3,5	3,5	3,5	0,0	-0,3%
Netherlands	3,5	3,5	3,4	3,5	3,4	-0,1	-2,3%
Belgium	2,9	2,9	2,7	3,0	3,0	0,0	1,1%
Brazil	3,2	3,5	3,5	3,3	2,9	-0,4	-12,0%
Switzerland	2,7	2,7	2,8	2,9	2,8	-0,1	-1,8%
Austria	2,7	2,8	3,0	2,4	2,4	0,0	2,0%
Serbia	2,3	2,3	2,4	2,4	2,4	0,0	-0,9%
Sweden	2,3	2,4	2,3	2,4	2,3	-0,1	-3,3%
Greece	3,1	3,0	2,6	2,4	2,3	-0,1	-4,4%
Chile	3,2	2,9	3,0	2,1	2,2	0,1	4,8%
Hungary	2,0	1,9	2,2	2,2	1,9	-0,3	-12,7%
Denmark	1,5	1,6	1,6	1,6	1,6	0,0	0,0%
Croatia	1,4	1,4	1,2	1,1	1,2	0,1	6,7%
Poland	0,9	0,9	1,0	1,1	1,1	0,1	4,9%
Bulgaria	1,0	0,8	0,9	1,0	1,0	0,0	3,4%
World total	244	243	240	241	242	0,9	0,4%

Sources: OIV, OIV Experts, Trade Press

a) Countries for which information has been provided with wine consumption of more than 1 mhl

b) 2015: provisional data

c) 2016: forecasted data

* Apparent consumption calculated by "Production + Imports - Exports" data for 2015 and 2016

Source: Organisation Internationale de la Vigne et du Vin, 2016

Min-Max

A simple alternative:

$$I_{q,c}^t = \frac{x_{q,c}^t - \min_{c,t}(x_q)}{\max_{c,t}(x_q) - \min_{c,t}(x_q)}$$



where the minimum and maximum for each indicator are calculated across countries and time. The normalized indicators have values between 0 and 1.

When **data for a new time point** become available the global minimum and/or the maximum may be affected. To maintain comparability between the existing and the new data, the composite indicator for the existing data must be re-calculated.

Min-Max

The expression

$$I_{q,c}^t = \frac{x_{q,c}^t - \min_c(x_q^{t_0})}{\max_c(x_q^{t_0}) - \min_c(x_q^{t_0})}$$

is sometimes used in time-dependent studies. However, if:

$$x_{q,c}^t > \max_c(x_q^{t_0})$$

the normalised indicator would be larger than 1

Categorical scale: original score and rescaling

Country	High computer skills	Categorical Scale	th Percent
FI	46.00	100	100-90
LU	42.00	100	100-90
DK	39.00	80	75-90exl
SE	38.00	80	75-90exl
EE	37.00	80	75-90exl
AT	34.00	60	50-75exl
LT	34.00	60	50-75exl
PT	34.00	60	50-75exl
ES	33.00	60	50-75exl
FR	33.00	60	50-75exl
UK	33.00	60	50-75exl
SI	31.00	60	50-75exl
DE	30.00	40	25-50exl
EL	30.00	40	25-50exl
LV	30.00	40	25-50exl
CZ	27.00	40	25-50exl
NL	27.00	40	25-50exl
IE	26.00	40	25-50exl
IT	26.00	40	25-50exl
MT	26.00	40	25-50exl
SK	25.00	40	25-50exl
BE	24.00	20	10-25exl
HU	24.00	20	10-25exl
CY	23.00	20	10-25exl
HR	22.00	20	10-25exl
PL	21.00	0	0-10exl
BG	15.00	0	0-10exl
RO	7.00	0	0-10exl

