

Representativeness of Knowledge Bases with the Generalized Benford's Law

Arnaud Soulet, Arnaud Giacometti, Béatrice Markhoff and Fabian M. Suchanek

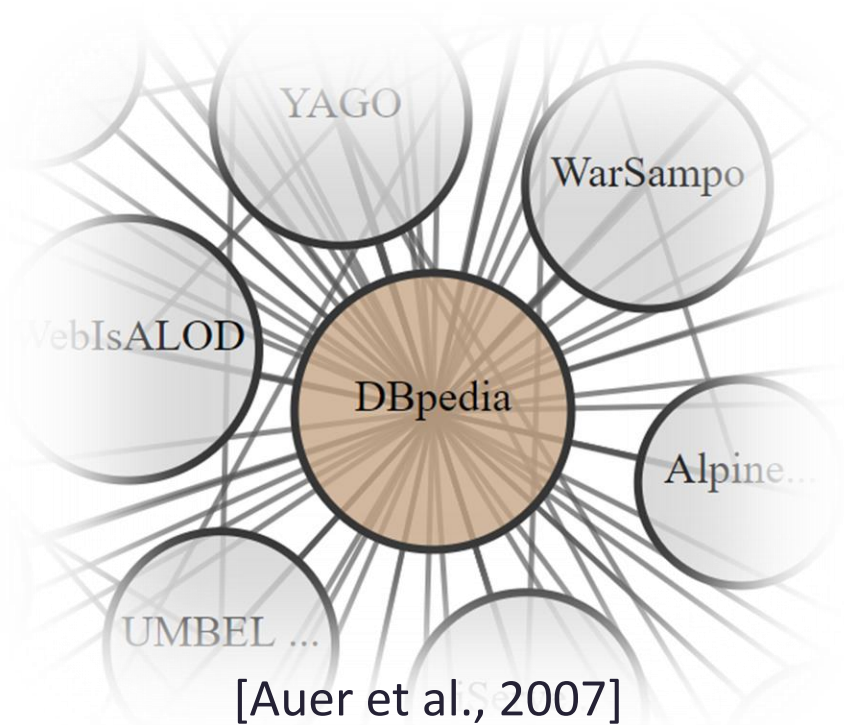
University of Tours



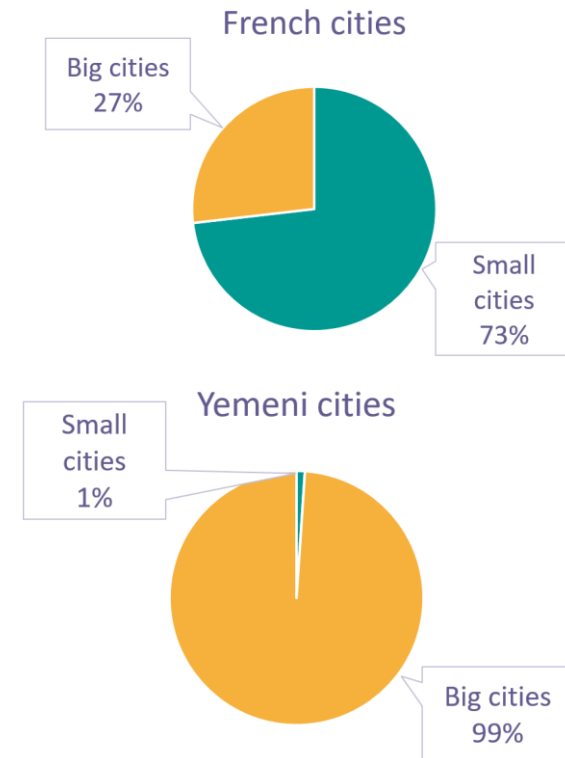
Telecom ParisTech



Reliability of queries on Knowledge Bases

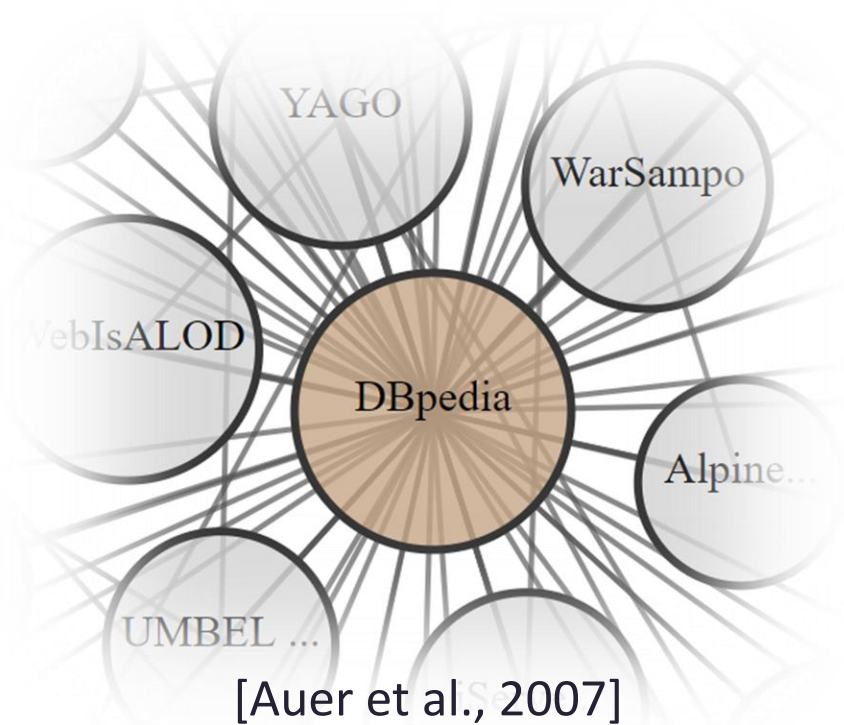


statistical
query

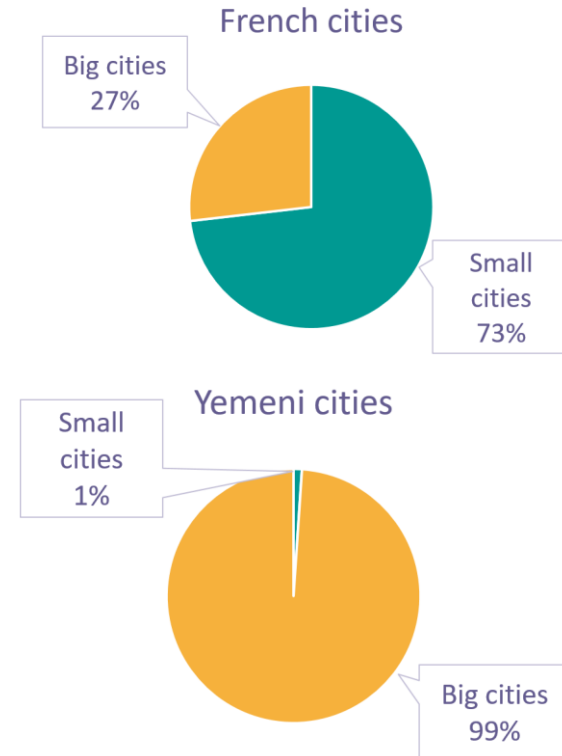


How many cities are small (<1k inhabitants) in France/Yemen?

Reliability of queries on Knowledge Bases



statistical
query

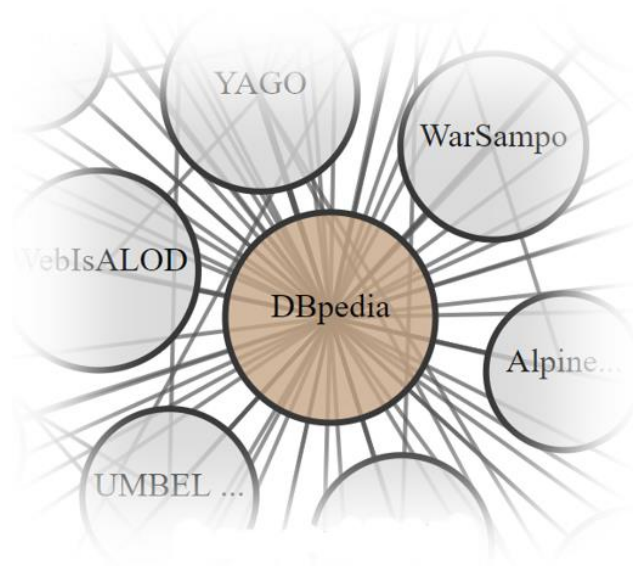


Does Yemen really not have any small cities?

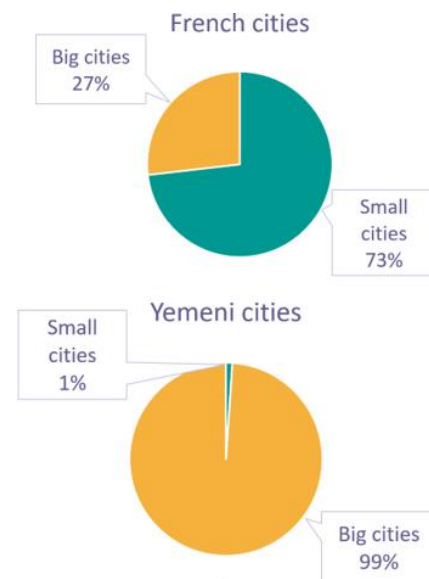
Reliability of queries on Knowledge Bases



crowdsourcing



statistical
query



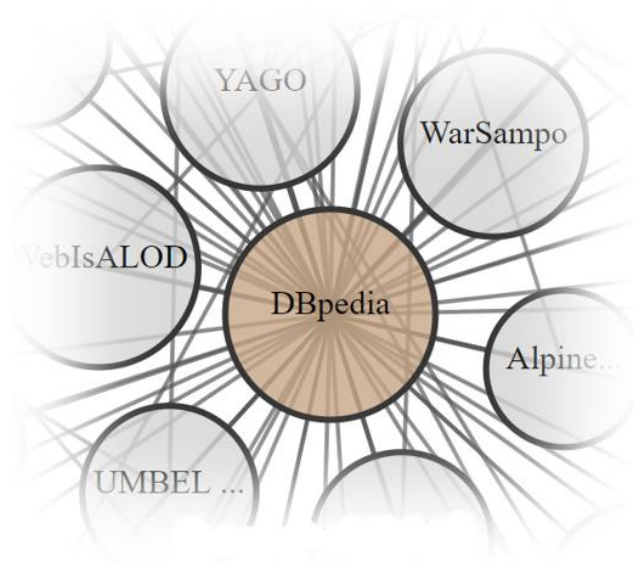
Voluntary bias

[Callahan and Herring, 2011;Wagner et al., 2015]

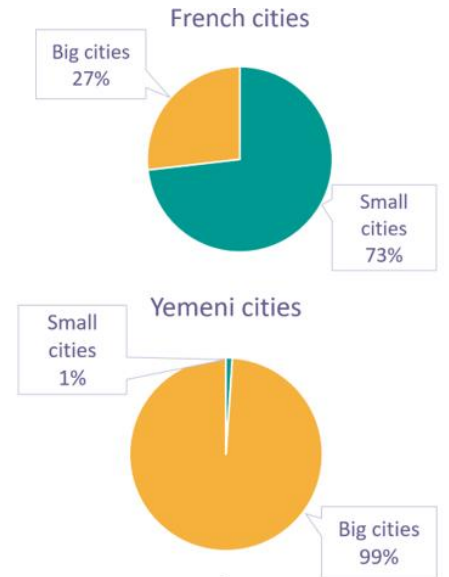
Reliability of queries on Knowledge Bases



crowdsourcing



statistical
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Voluntary bias

[Callahan and Herring, 2011;Wagner et al., 2015]

We do not know the KB biases, but the statistics can give us a hint?

Missing facts



Yemeni
city:

Sanaa

Population:

1,937,451



Aden

760,923



Taiz

missing
615,222

[Darari et al., 2016;
Galarraga et al., 2017;
Lajus and Suchanek, 2018;
Razniewski et al., 2015;
Razniewski et al., 2016]

Several methods for estimating the completeness for facts

Missing facts $\neq$ Missing entities + missing facts



Yemeni
city:

Sanaa

Population:

1,937,451



Aden

760,923



Taiz

missing
615,222



Haid al-
Jazil
missing

missing
few

Missing facts due to missing entities are ignored!

Completeness

$$= \text{\#present facts} / (\text{\#present facts} + \text{\#missing facts})$$

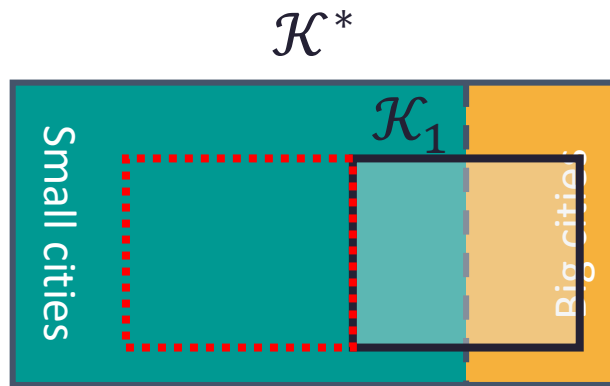
Assuming that $\mathcal{K}^*$ is an ideal KB (= correct + complete):



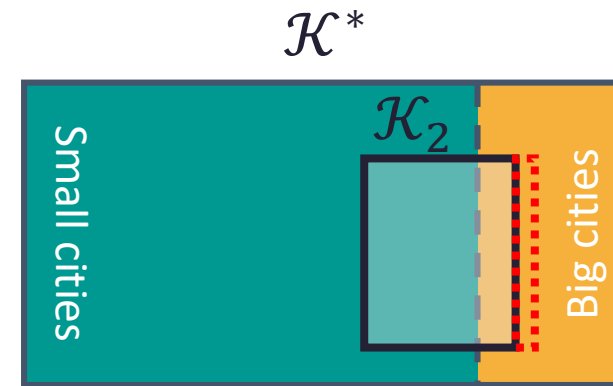
What is the best KB between $\mathcal{K}_1$ and $\mathcal{K}_2$ for statistical queries?

Completeness $\neq$ Representativeness

Assuming that $\mathcal{K}^*$ is an ideal KB (= correct + complete):



More complete, less representative



Less complete, more representative



Representativeness is more important than completeness for statistics!

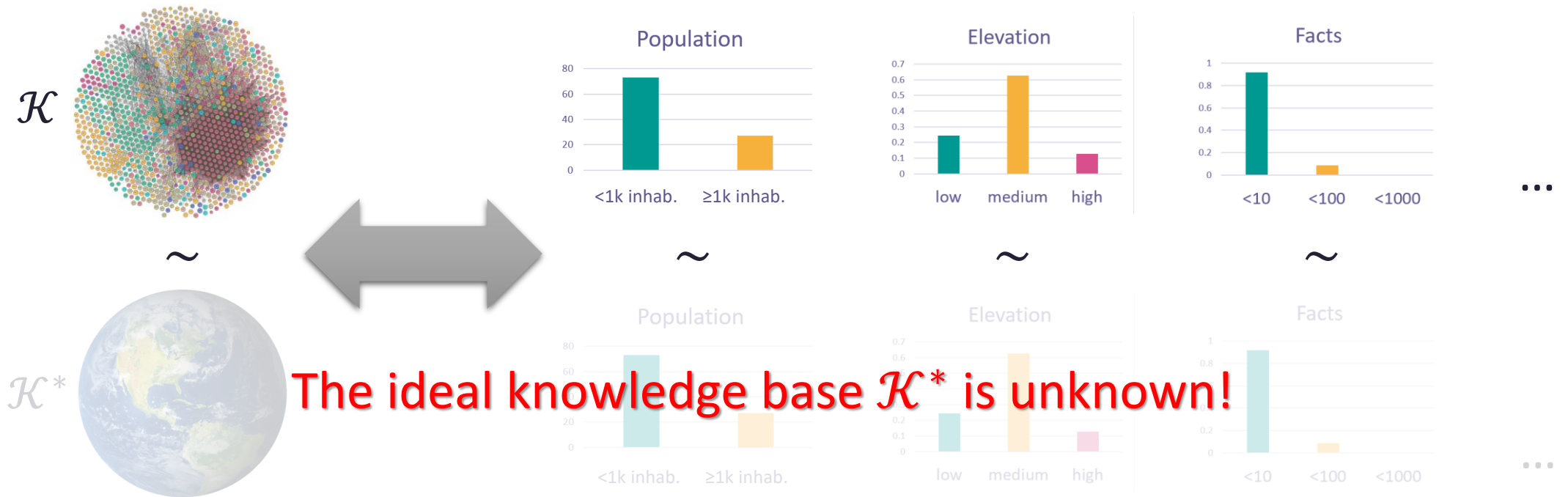
Representativeness of Knowledge Bases

A KB $\mathcal{K}$ is representative of $\mathcal{K}^*$ iff the distribution remains the same for all uniform-sampling invariant measures.



Representativeness of Knowledge Bases

A KB $\mathcal{K}$ is representative of $\mathcal{K}^*$ iff the distribution remains the same for all uniform-sampling invariant measures.



Challenge: How to estimate the representativeness?

Example: population of capitals

Abidjan	Bangkok	Conakry	Kingston	Mogadishu	Santiago
Abuja	Beijing	Dakar	Kinshasa	Montevideo	Seoul
Accra	Belgrade	Damascus	Kuala Lumpur	Nairobi	Sofia
Addis Ababa	Berlin	Dhaka	Lilongwe	Niamey	Taipei
Algiers	Bogota	Doha	Lima	Ouagadougou	Tashkent
Amman	Brasilia	Erbil	London	Paris	Tbilisi
Ankara	Brazzaville	Freetown	Luanda	Phnom Penh	Tegucigalpa
Antananarivo	Bucharest	Havana	Lusaka	Prague	Tokyo
Ashgabat	Budapest	Islamabad	Madrid	Pyongyang	Tripoli
Bahawalpur	Buenos Aires	Jakarta	Managua	Quito	Tunis
Baku	Cairo	Kabul	Maputo	Riyadh	Ulaanbaatar
Bamako	Caracas	Khartoum	Mexico City	Sana'a	Vienna

Example: population of capitals

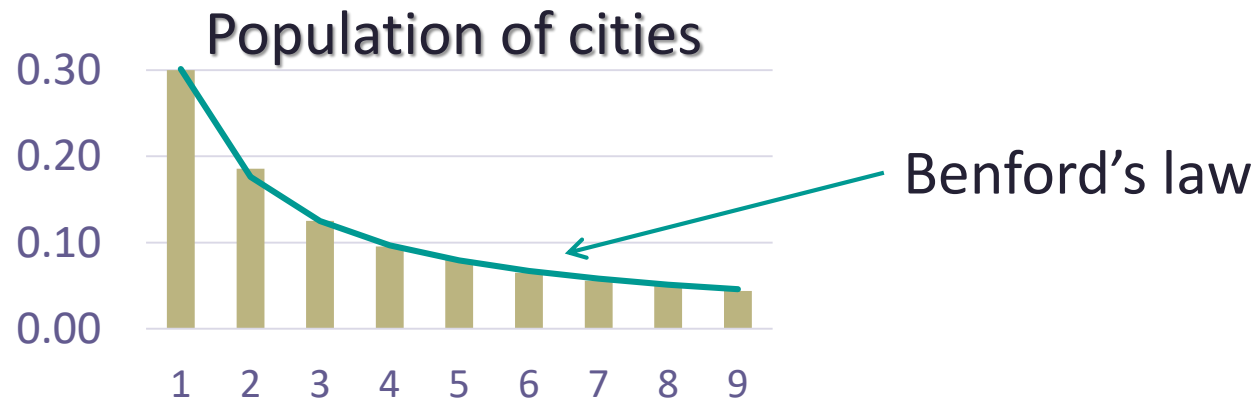
4 707 404	8 280 925	1 660 973	1 041 084	1 750 000	6 158 080
1 235 880	21 700 000	1 146 053	10 125 000	1 305 082	9 971 111
2 291 352	1 166 763	1 711 000	1 768 000	3 138 369	1 260 120
3 384 569	3 610 156	6 970 105	1 077 116	1 302 910	2 704 974
3 415 811	7 878 783	1 351 000	8 852 000	1 626 950	2 309 600
4 007 526	2 556 149	1 025 000	8 673 713	2 229 621	1 118 035
4 587 558	1 827 000	1 050 301	2 825 311	1 501 725	1 157 509
1 613 375	1 883 425	2 106 146	1 742 979	1 267 449	13 617 445
1 031 992	1 759 407	1 900 000	3 141 991	2 581 076	1 126 000
1 052 000	2 890 151	9 607 787	2 205 676	2 671 191	1 056 247
2 122 300	10 230 350	3 678 034	1 766 184	7 125 180	1 372 000
1 809 106	3 273 863	5 185 000	8 918 653	1 937 451	1 852 997

Example: population of capitals

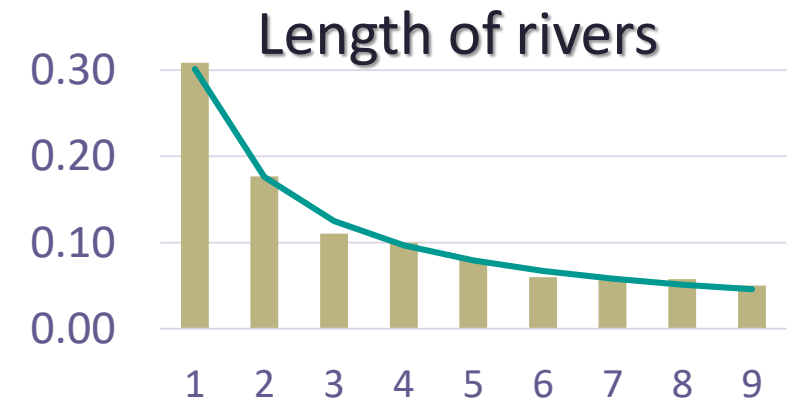
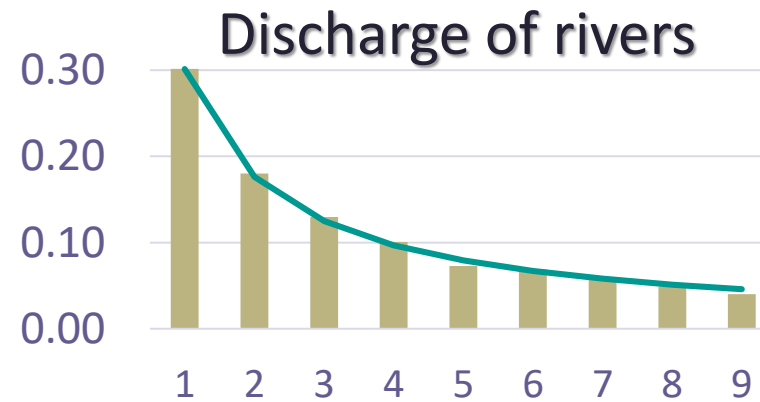
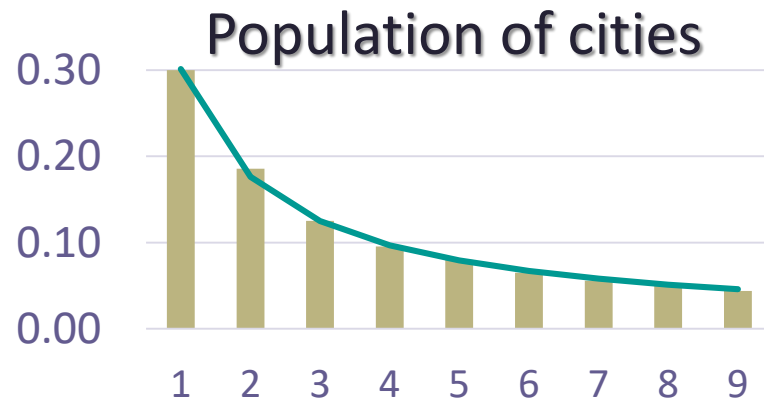
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What is the distribution of the first significant digit of capital inhabitants?

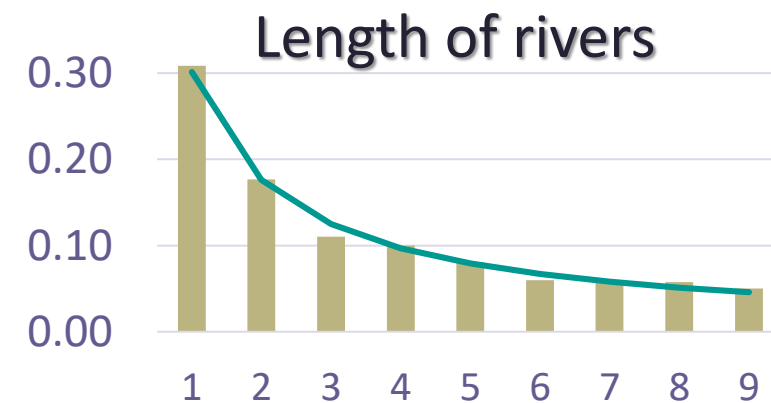
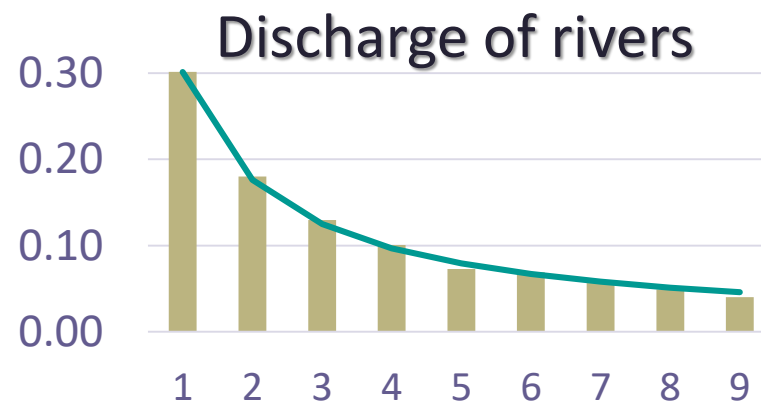
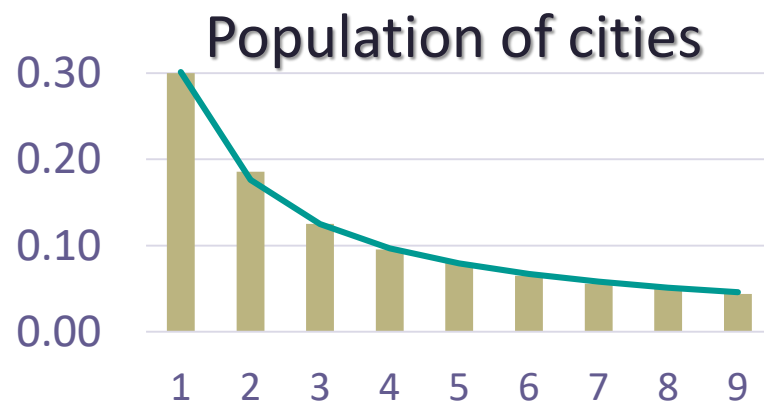
Benford's law



Benford's law



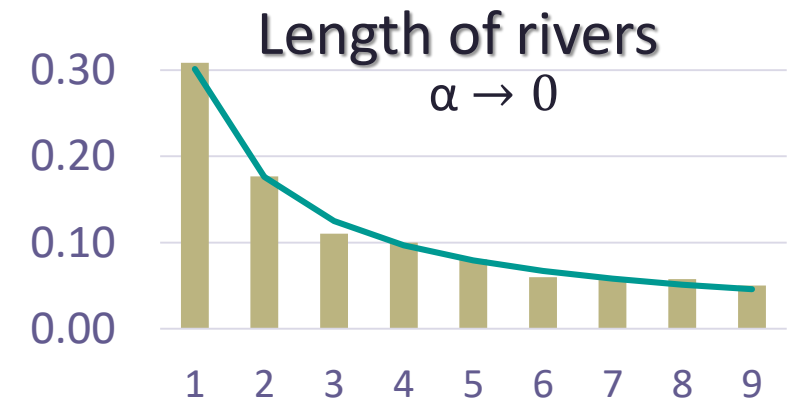
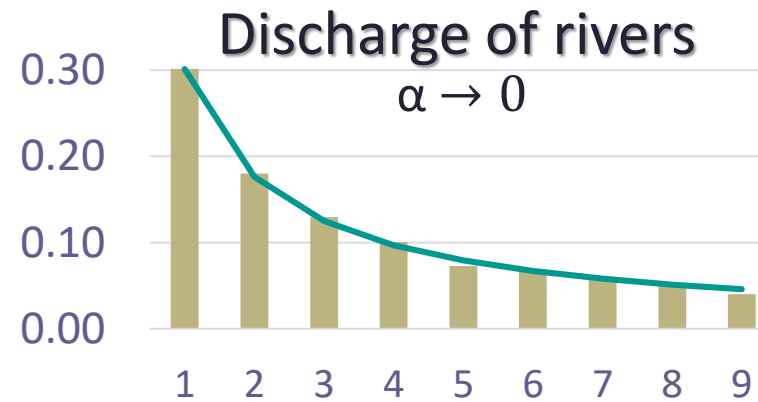
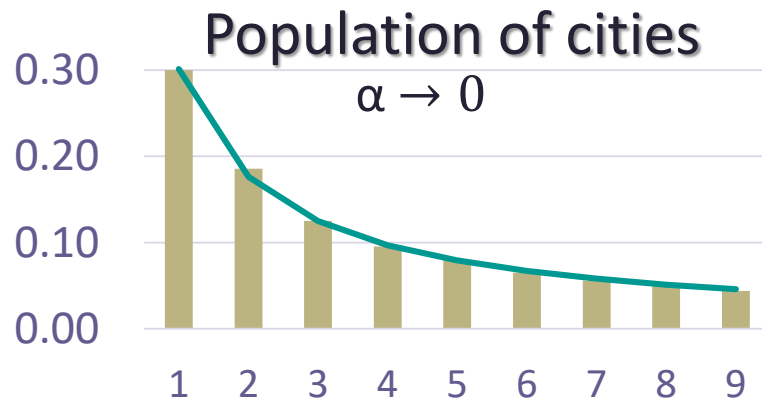
Benford's law



$$P(\text{first digit}(X) = d) = \log \left(1 + \frac{1}{d} \right)$$

[Newcomb, 1881; Benford, 1938]

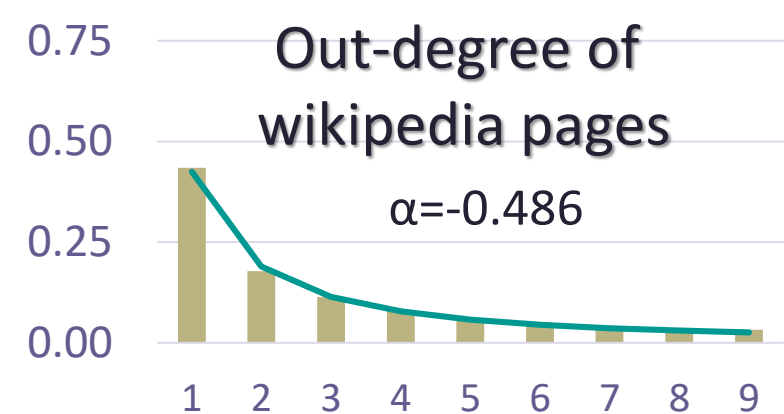
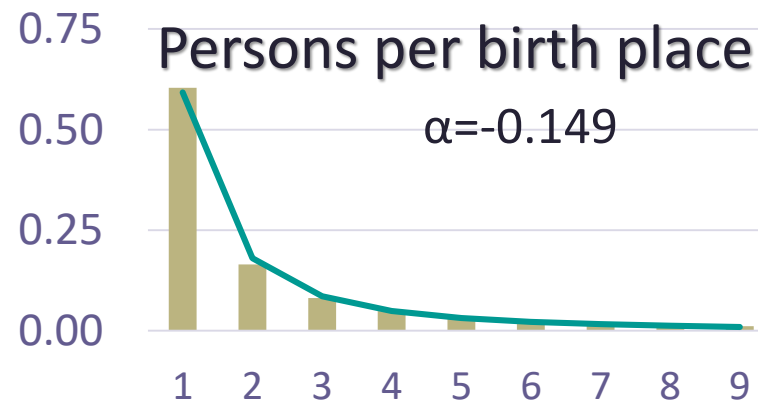
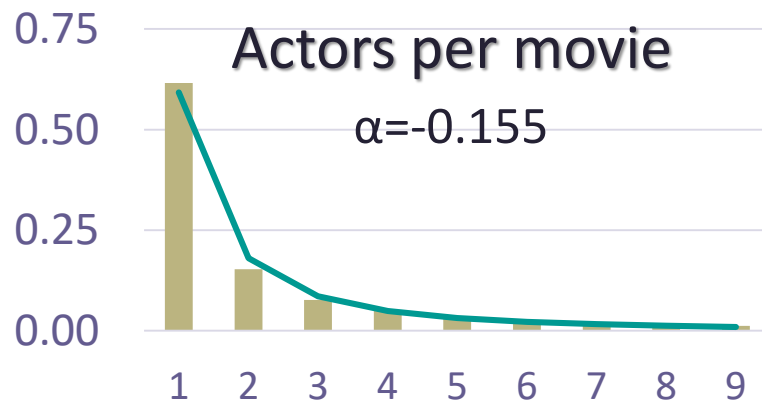
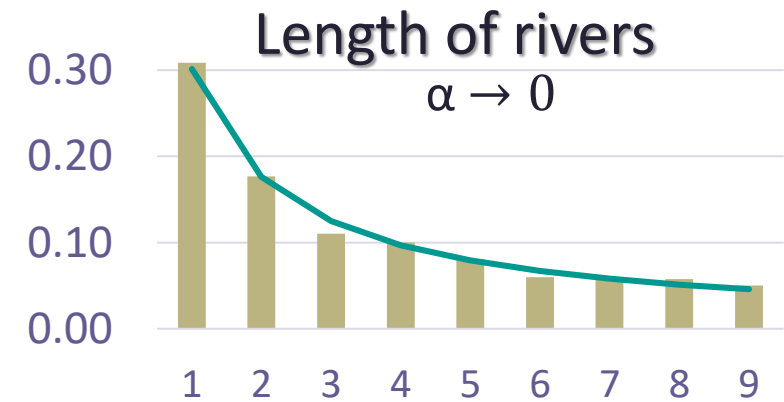
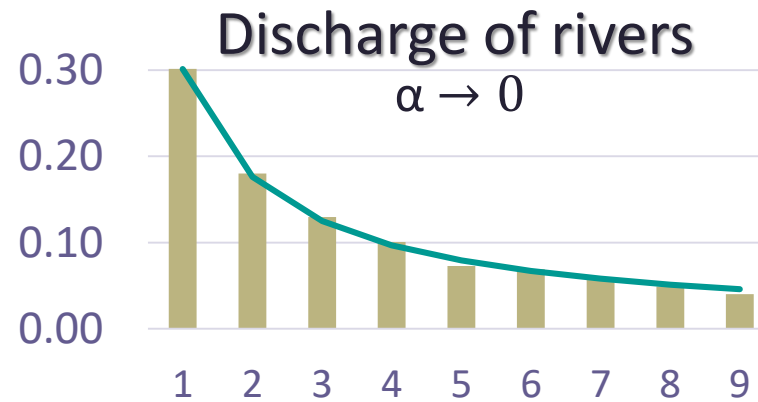
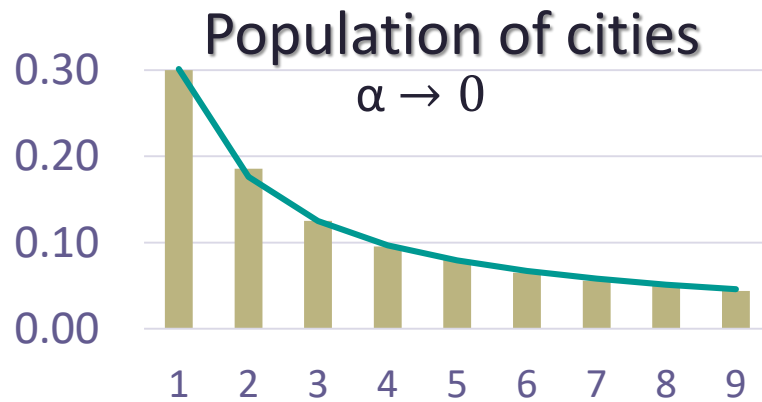
The Generalized Benford's Law



$$P(\text{first digit}(X) = d) = \frac{(1 + d)^\alpha - d^\alpha}{10^\alpha - 1}$$

[Hürlimann, 2014]

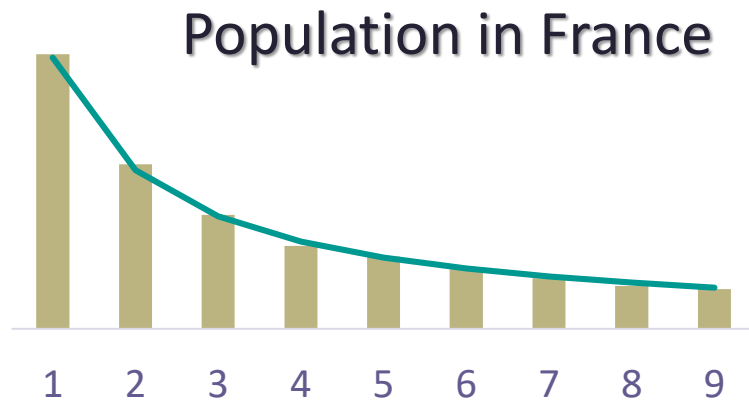
The Generalized Benford's Law



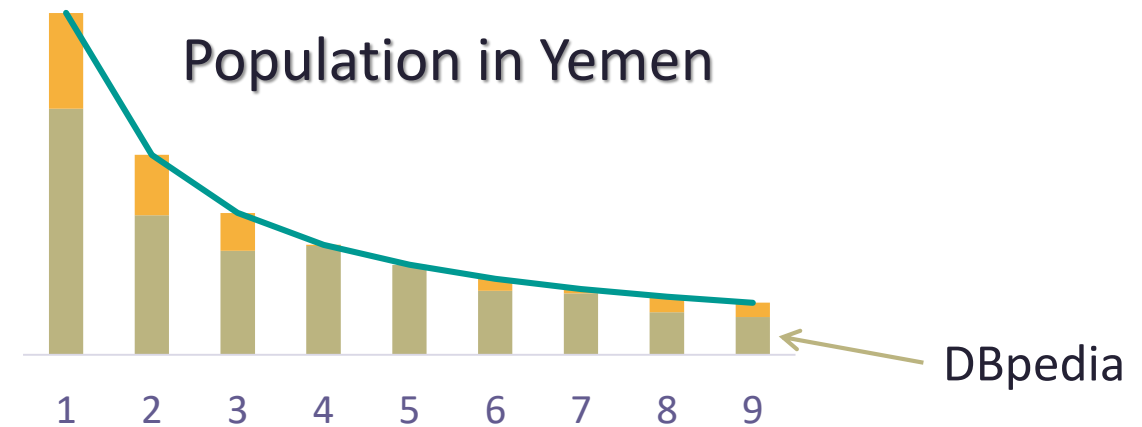
Key idea of our method

representativeness = compliance with the Generalized Benford's Law

$$= \frac{\#present_facts}{\#present_facts + \#missing_facts_for_compliance}$$



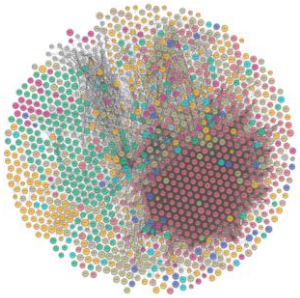
Representativeness = 97%



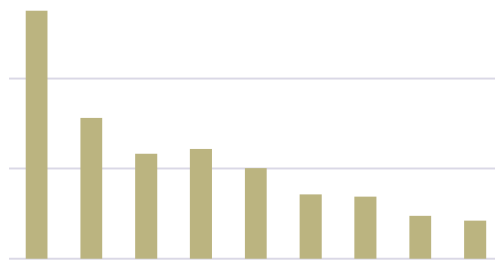
Representativeness = 79%

Our method in supervised context

facts of r on $\mathcal{K}$



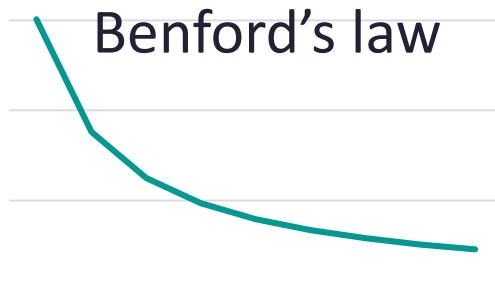
distribution of the fsd



facts of r on $\mathcal{K}^*$



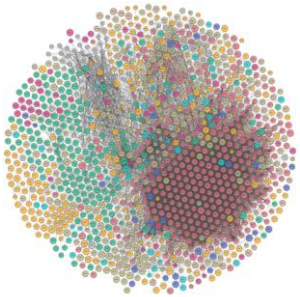
Benford's law



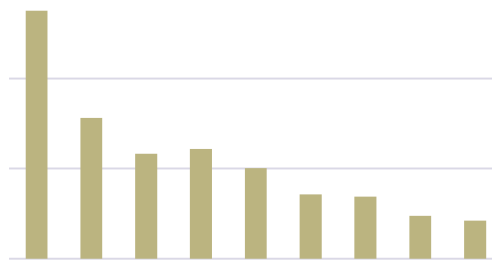
Using the known distribution of the first significant digit

Our method in supervised context

facts of r on $\mathcal{K}$



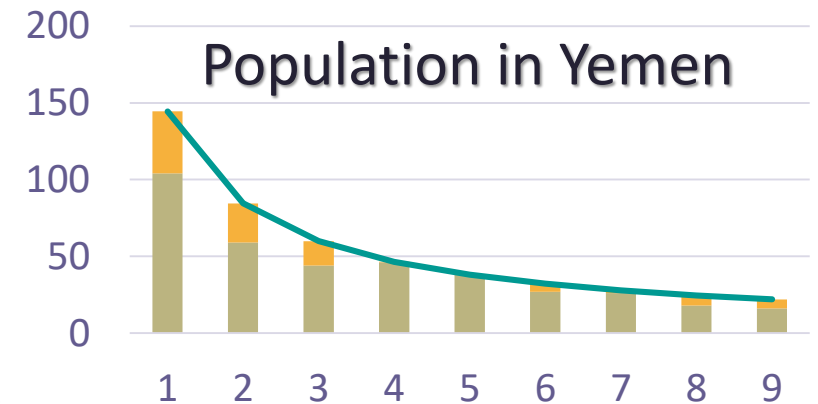
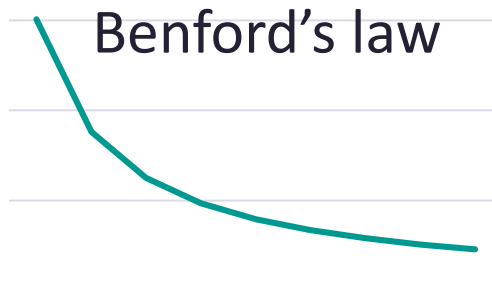
distribution of the fsd



facts of r on $\mathcal{K}^*$

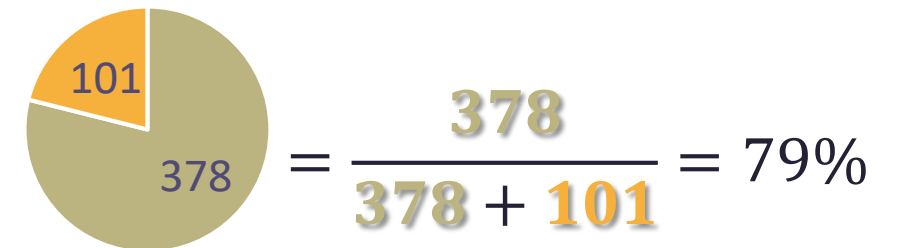


Benford's law



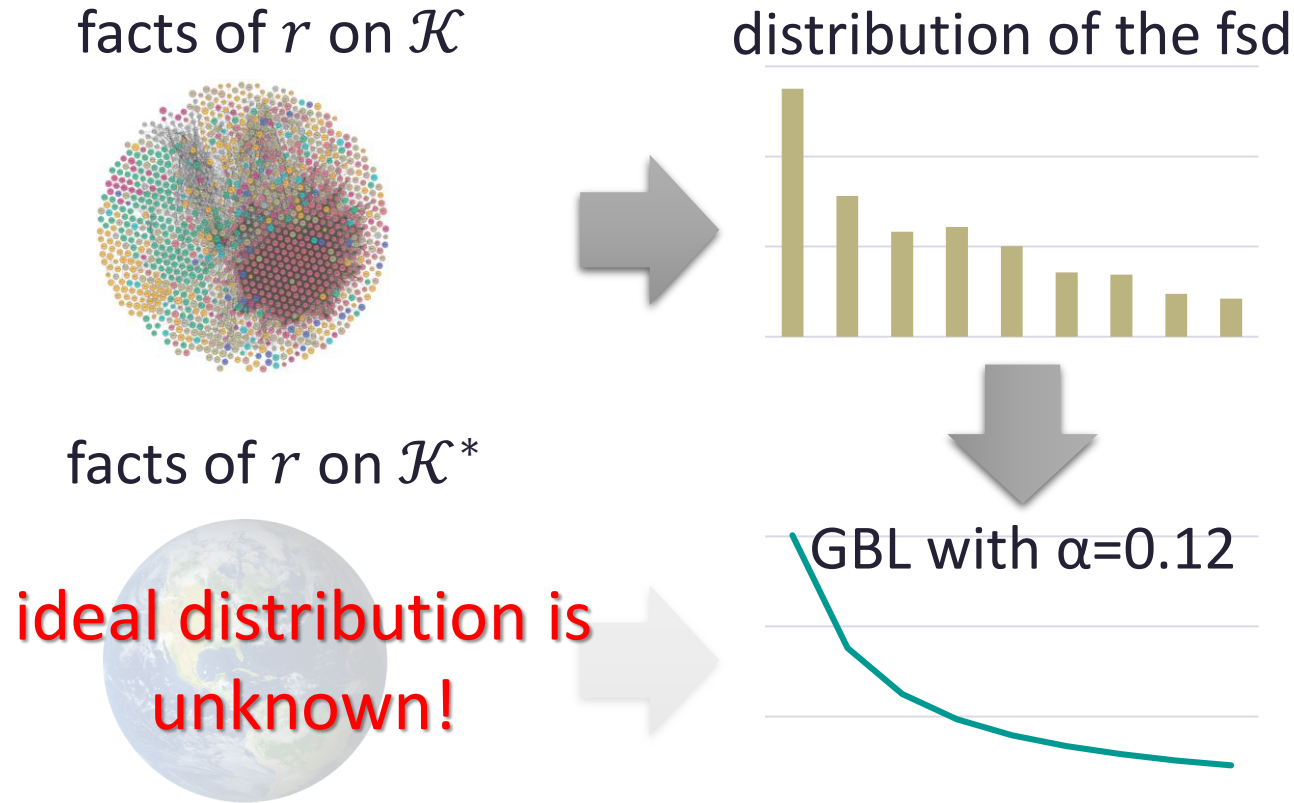
378 present facts 101 missing facts

Representativeness:



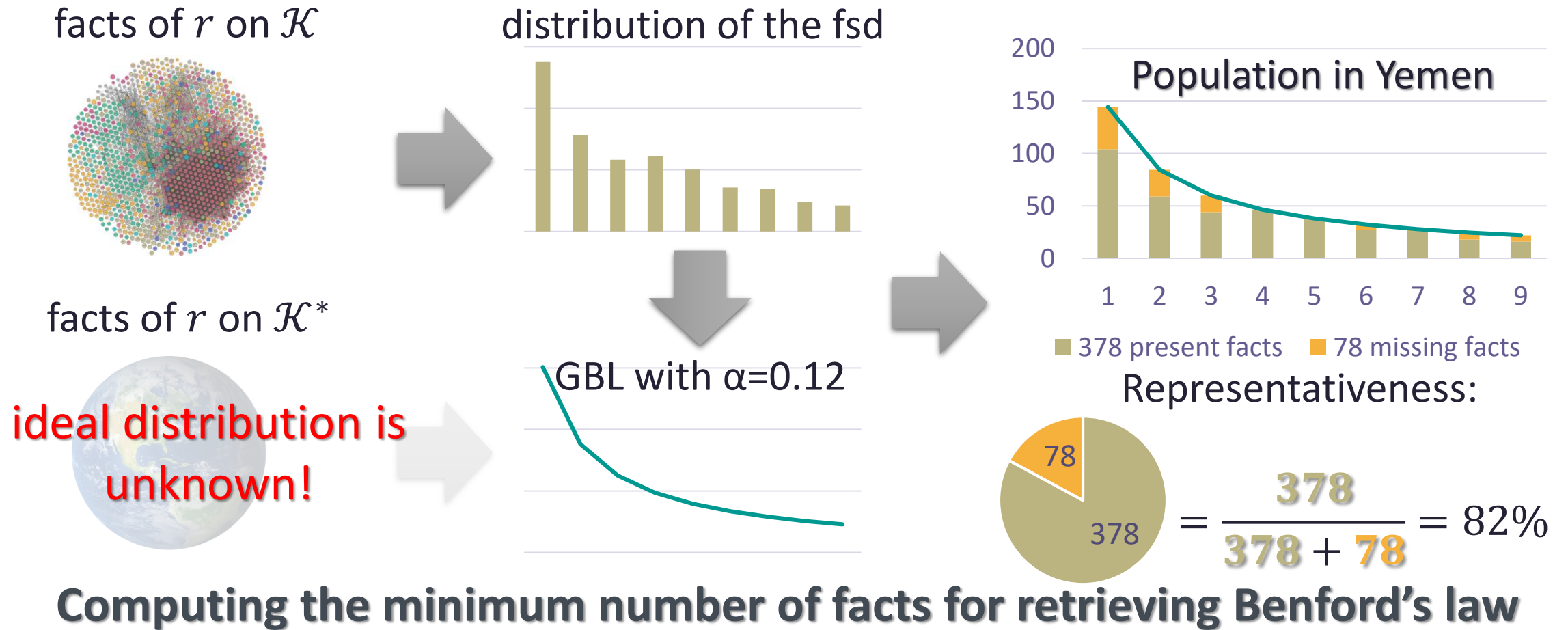
Computing the minimum number of facts for retrieving Benford's law

Our method in unsupervised context



Learning the parameter α of the Generalized Benford's Law

Our method in unsupervised context



Experimental study

❑ **Evaluation protocol**

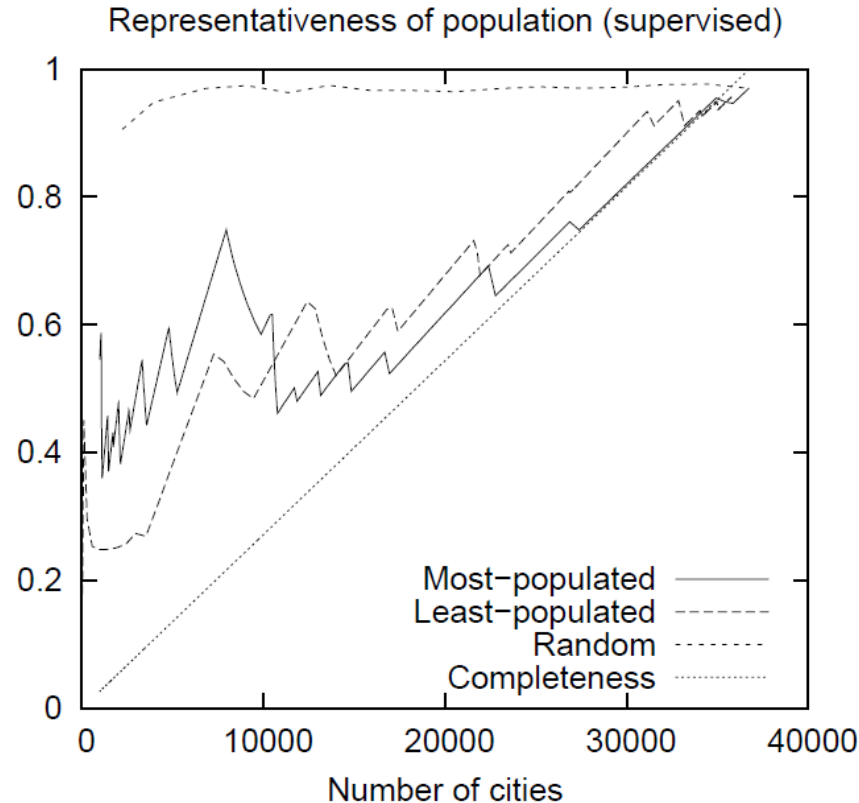
1. Take a correct and complete relation as gold standard
2. Degrade the completeness by discarding facts
3. Approximate the representativeness

❑ **Gold standard:** population in French cities according to govt statistics

❑ **Degradation:**

- Most-populated: remove the least populated cities
- Least-populated: remove the most populated cities
- Random: remove cities randomly

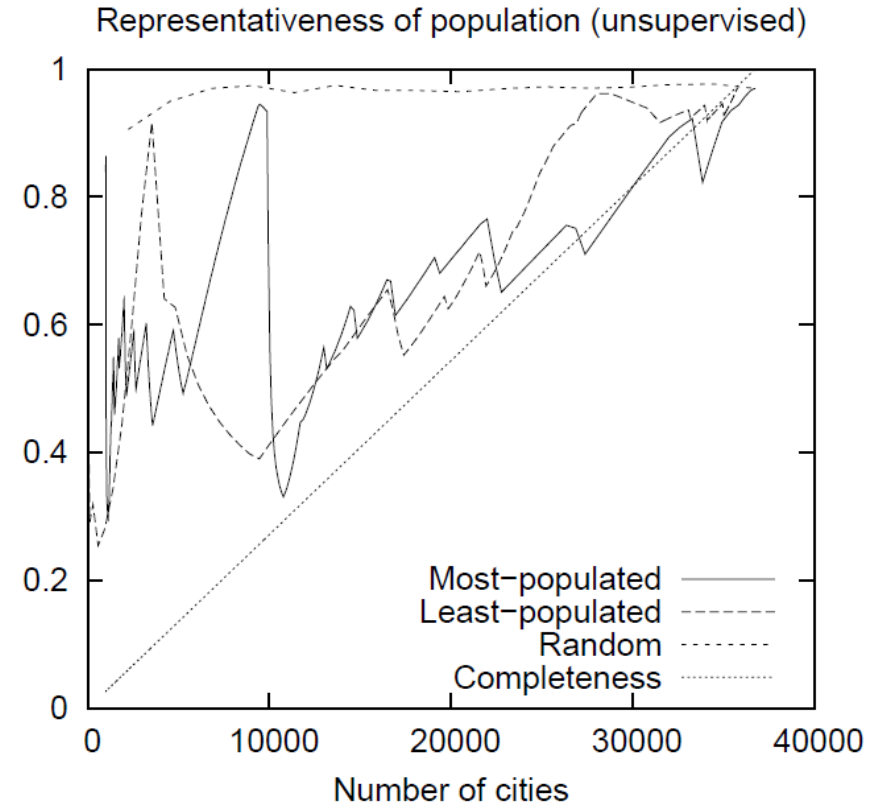
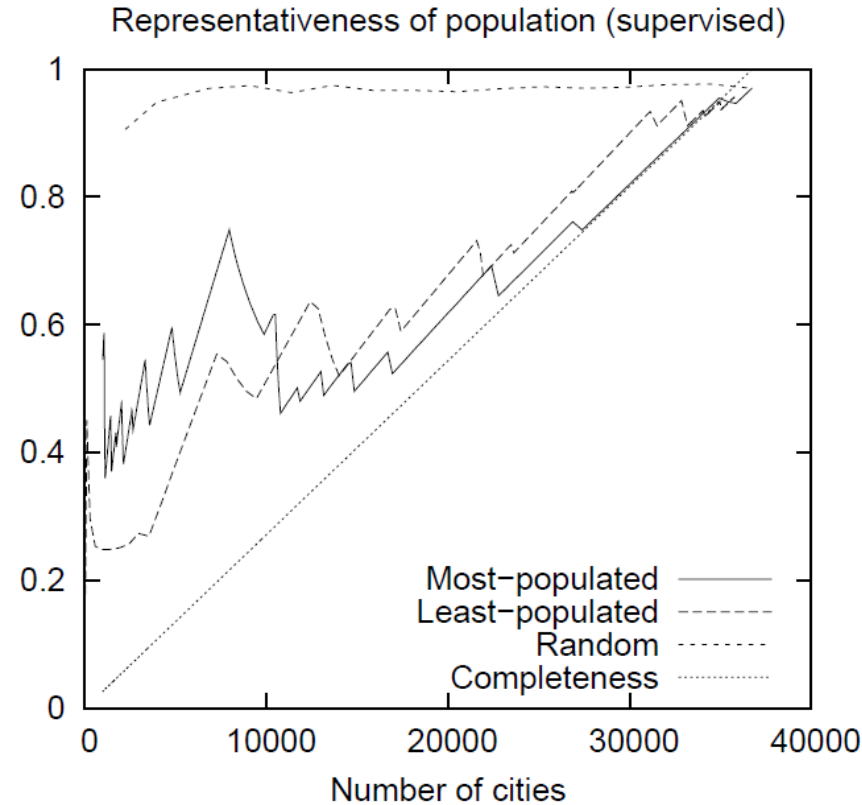
Population of French cities



- Representativeness is an upper bound of completeness
- **Most/least-populated degradation:** tight bound if number of cities > 22k
- **Random degradation:** the representativeness is high

Representativeness approximates well the bias

Population of French cities

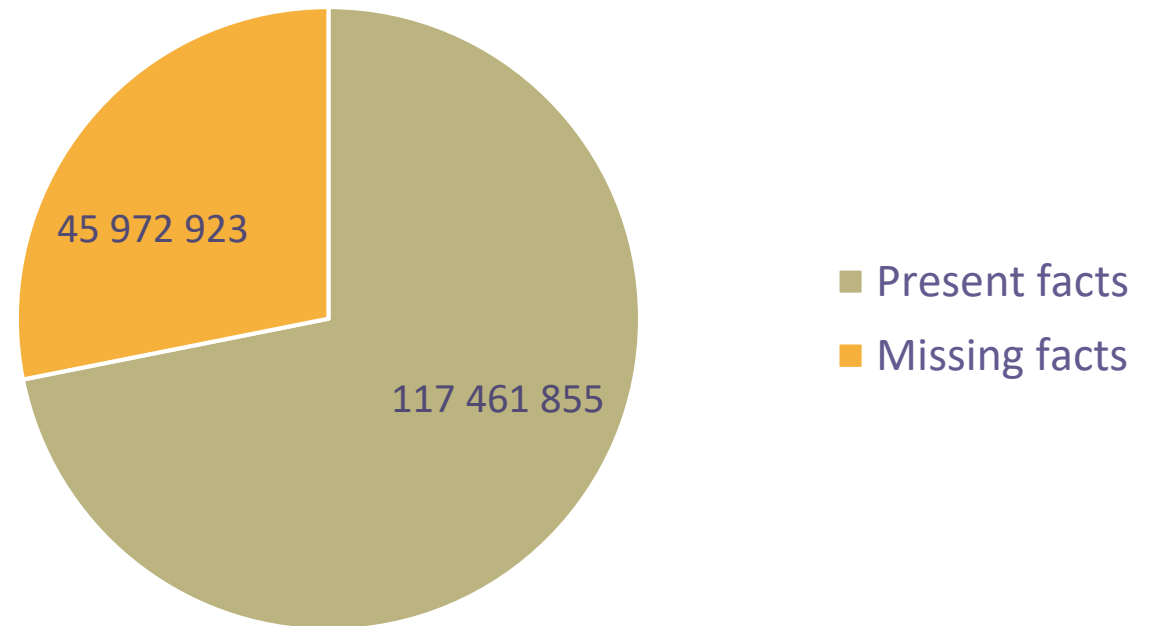


Learning the parameter α does not perturbate the approximation

Auditing DBpedia (France)

1,487 relations (out of 2,920) have a distribution statistically compliant with the GBL

Representativeness:
72%



Conclusion

- ❑ **The representativeness is more important than the completeness for achieving statistics.**
- ❑ **First use of Benford's law for approximating the proportion of missing data**
- ❑ **The approximate representativeness based on the GBL is an upper bound of the true representativeness and the true completeness.**
- ❑ **Future work:**
 - How to correct sparql queries with representativeness information?
 - How to scale up the approach to audit the LOD?