

Fake Content & Bias

- British Prime Minister Benjamin Disraeli:
 - "There are three kinds of lies: lies, damned lies, and statistics.

UTC professor says "Everyone has bias"











Here Are 50 Of The Biggest Fake News Hits On Facebook From 2016

One fake news entrepreneur says we should expect even more Trump hoaxes in

sted on Dec. 30, 2016, at 2:12 p.m

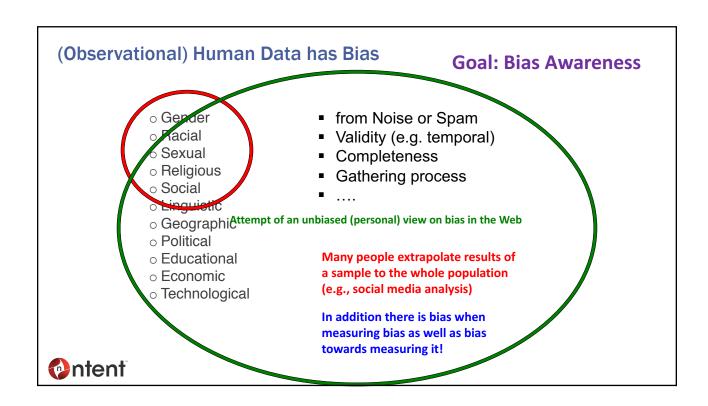


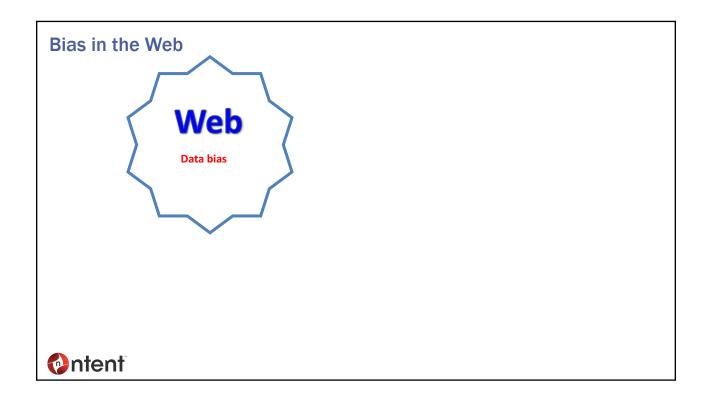
Craig Silverman BuzzFeed News Media Editor

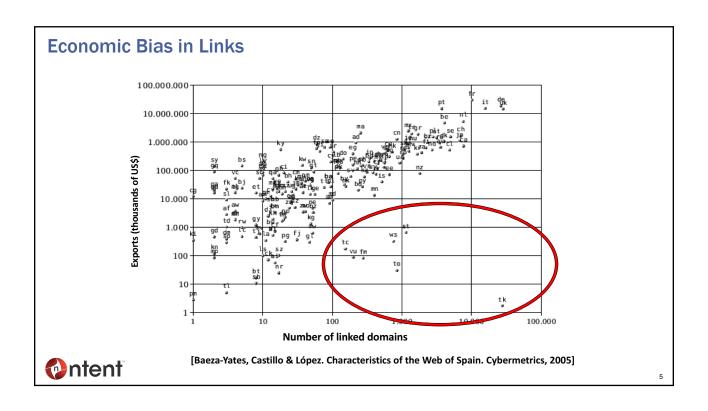
Bias: significant deviation from a prior (unknown) distribution

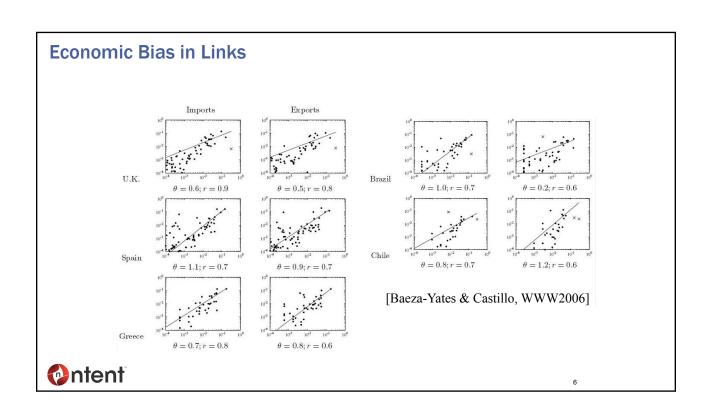
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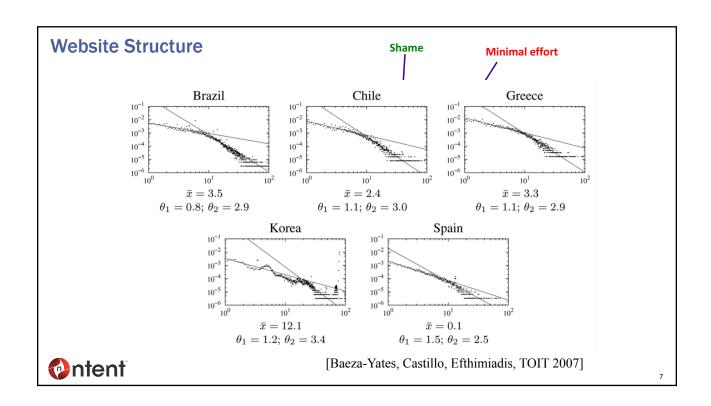
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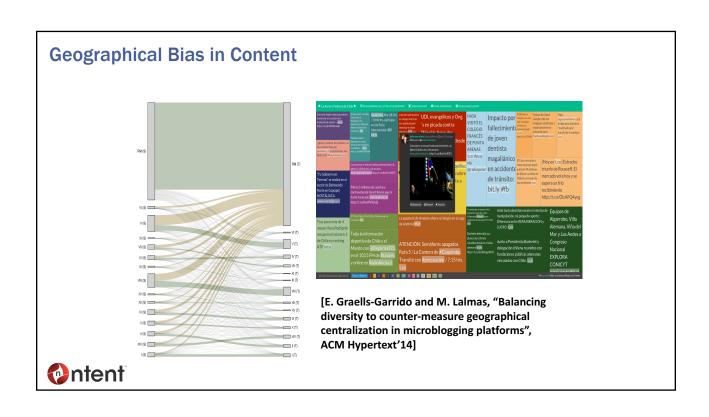












Gender Bias in Content

Word embedding's in w2vNEWS

Gender stereotype she-he analogies.

sewing-carpentry register-nurse-physician housewife-shopkeeper nurse-surgeon interior designer-architect softball-baseball blond-burly feminism-conservatism cosmetics-pharmaceuticals giggle-chuckle petite-lanky vocalist-guitarist diva-superstar charming-affable sassy-snappy volleyball-football cupcakes-pizzas hairdresser-barber

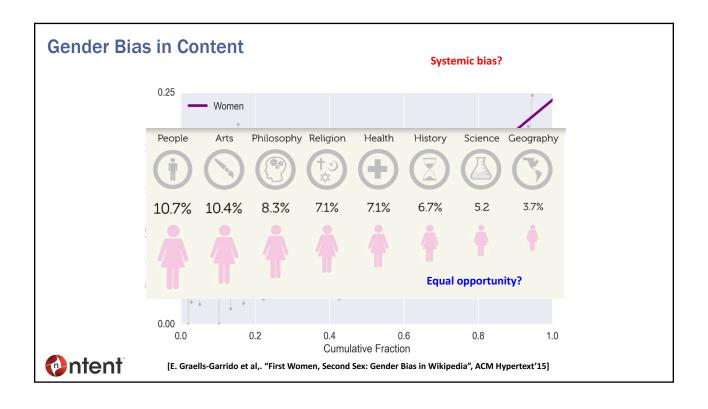
Gender appropriate she-he analogies.

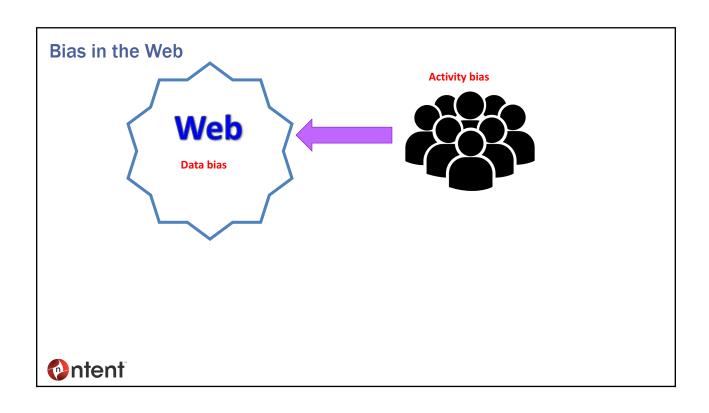
queen-king sister-brother mother-father waitress-waiter ovarian cancer-prostate cancer convent-monastery

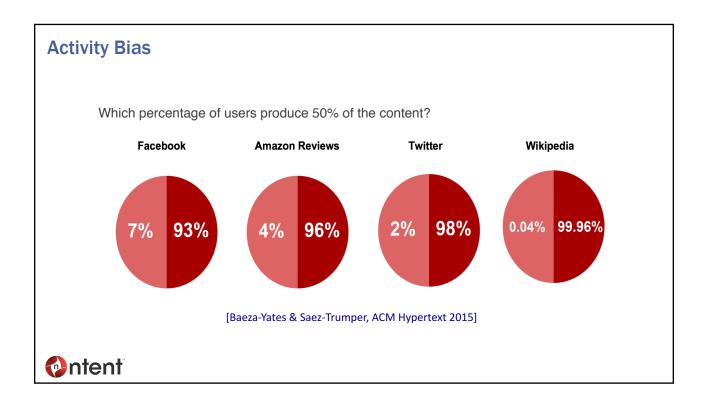
Most journalists are men?

[Bolukbasi at al, ArXiv 2016] Yes, about 60 to 70% at work although at college is the inverse





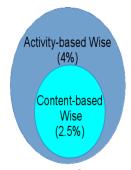


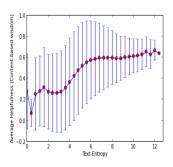




Quality of Content?

- Adding content implies adding wisdom?
- We used Amazon's reviews helpfulness and computed the text entropy
- · Content-based-wise users
- How many of those users are being paid?

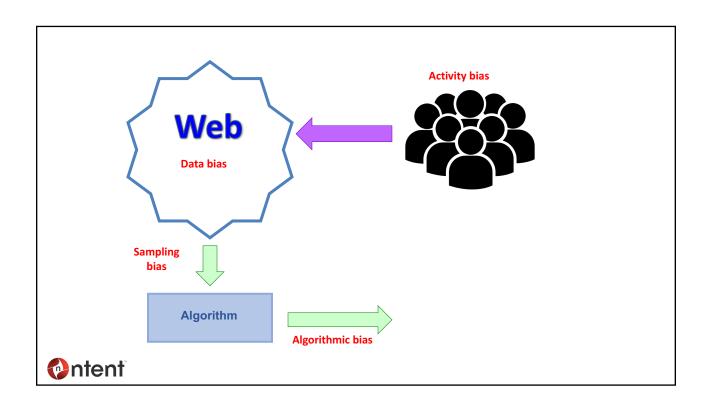






[Baeza-Yates & Saez-Trumper, ACM Hypertext 2015]

1.1% of the Twitter content is never seen.* 31% of articles added/edited in May 2014 in wikipedia, were not visited in June. (Baeza-Yates & Saez-Trumper, ACM Hypertext 2015)



Sample Size?

- If we want to estimate the frequency of queries that appear with probability at least p with a certain relative error ϵ we can use the standard binomial error formula $\sqrt{(1-p)/np}$ which works well for p near $\frac{1}{2}$ but not for p near 0
- Better is the Agresti-Coull technique (also called take 2) which gives:

$$n \ge Z_{1-\alpha/2}^2 \left(\frac{p'(1-p')}{\epsilon^2} - 1 \right)$$

where Z is the inverse of the standard normal distribution, $1-\alpha$ is the confidence interval and $\ p'=p+Z^2/2$

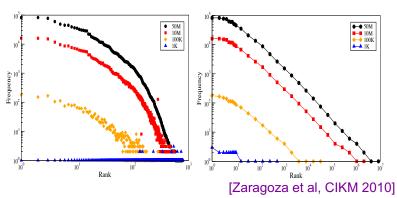
• If p = 0.1, $1 - \alpha$ is 90% and ϵ is 10%, we get n = 2342. The standard formula gives n = 900!



[Baeza-Yates, SIGIR 2015, Industry track]

Sampling Techniques

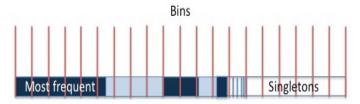
- $\bullet \quad \text{Standard technique:} \qquad p_q \approx \widehat{p}_q(\mathcal{S}) = \frac{f_q(\mathcal{S})}{\sum_{q' \in \mathcal{S}} f_{q'}(\mathcal{S})}$
- A good sample should cover well all the query distribution but this does not work with very biased distributions.





Incremental Stratified Sampling

- Main goal: make good samples consistent across time
- · Simple idea based in stratified sampling: bins + random start point



- Bin size can be found by binary search starting with a good approximation if a query frequency model is used (b < V/n)
- · This perfectly mimics the head of the distribution, but not the tail
- Change the bins in the tail to get the right distribution



[Baeza-Yates, SIGIR 2015, Industry track] 20

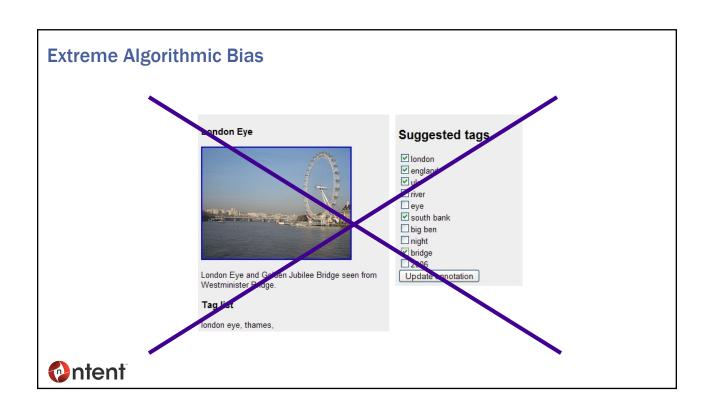
Stratified Sampling Example

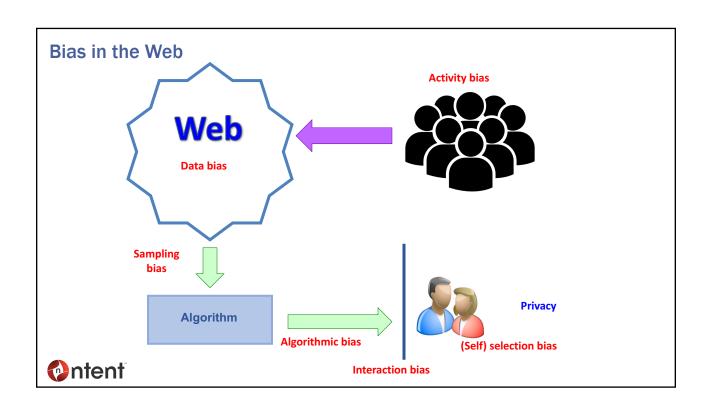
Output

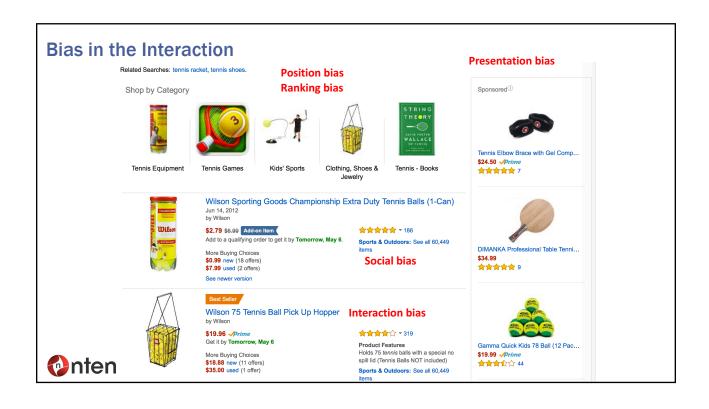
Description:

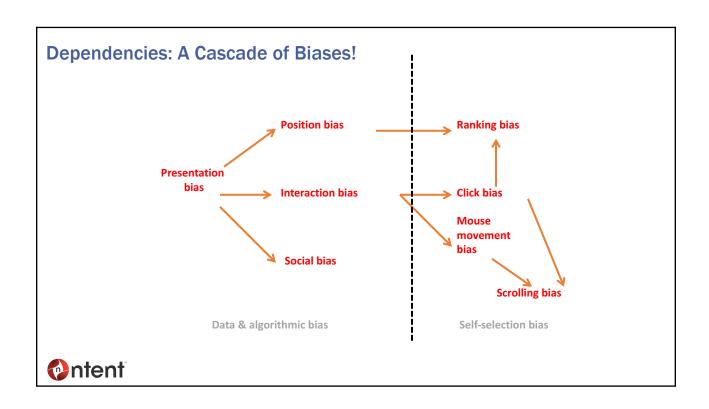
Output

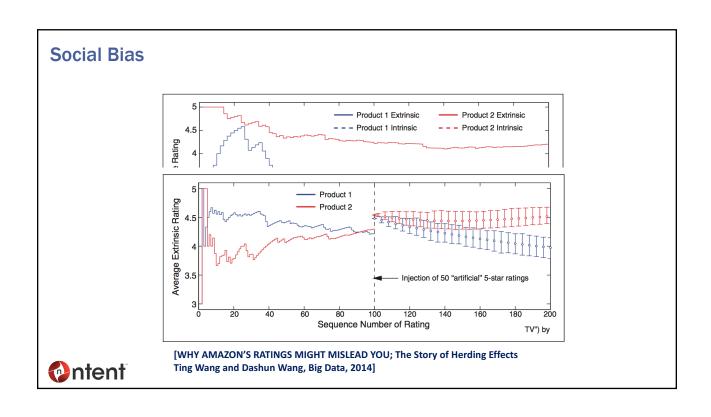
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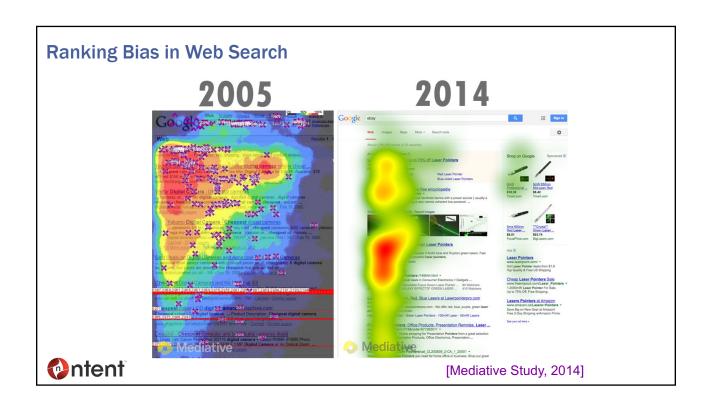




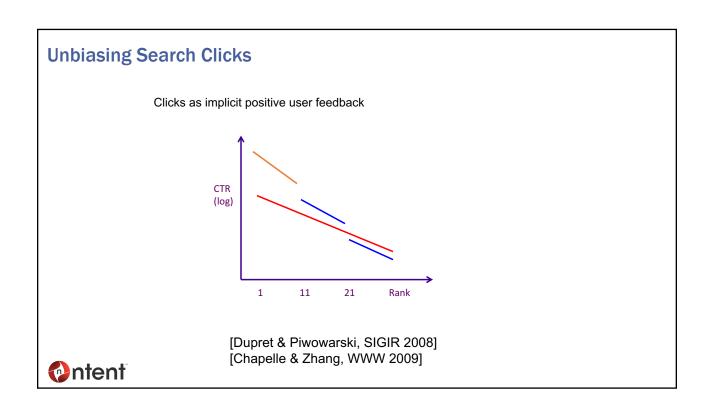


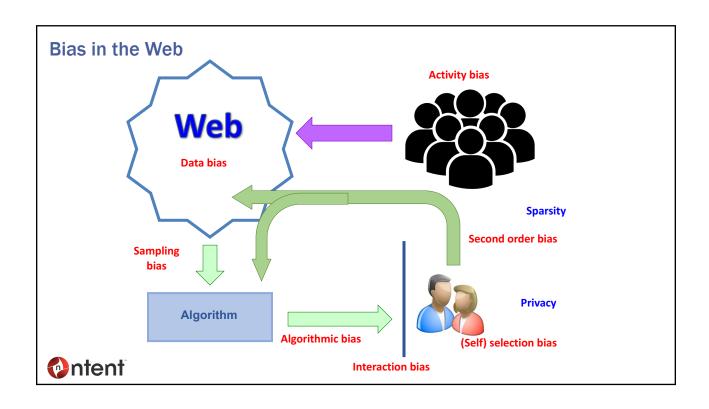


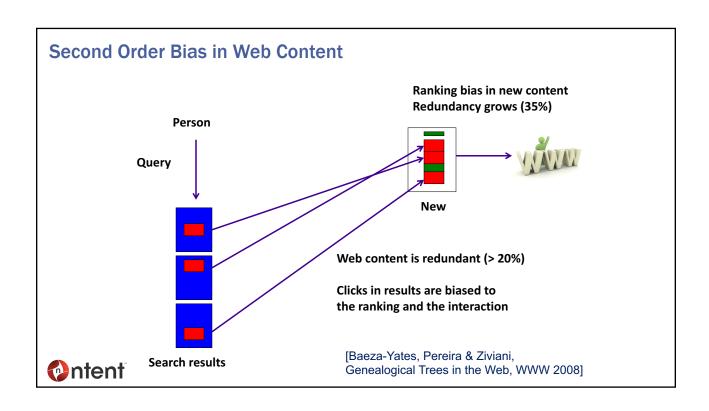












Avoid Second Order Bias due to Personalization

The Filter "Bubble", Eli Pariser (2011)

- The effect of self selection bias
- Avoid the poor get poorer syndrome
- · Avoid the echo chamber
- Empower the tail

Partial solutions:

- Diversity
- Novelty
- Serendipity
- · Show me the dark side

Cold start problem solution: Explore & Exploit

How much exploration is needed for presentation bias?





Aggregating in the Tail

Exploit the context (and deep learning!)

91% accuracy to predict the next app you will use [Baeza-Yates et al, WSDM 2015]

Personalization vs. Contextualization
 Recall that user interaction is another long tail

