## What makes two "languages" different?

Issues analyzed in Kleinberg (2004, Data Stream Management 2016), with a Markov model applied for temporal analysis.
Presentation/figures follow Monroe, Colaresi and Quinn, Political Analysis (2008)

## Persuasion: frame competition

Example: public discussion of GMOs in food

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```
"green revolution"
```


## Persuasion: frame competition

Example: public discussion of GMOs in food


## Additional applications:

 Differentiating the language of ....- successful vs. unsuccessful persuaders
- language in one time period vs. another...
- males vs females
- your experimental condition A vs. your experimental condition B!!

Also good for sanity-checking your data...

## Example: 106th U.S. Senate speeches on abortion

"Frames" $\rightarrow$ words we might expect from Democrats:

## women's rights

privacy
"Frames" $\rightarrow$ words we might expect from Republicans:

- Assume a joint vocabulary of terms $v_{i}$.
$p\left(v_{i}\right)$ and $p\left(v_{i}\right)$ : observed relative frequency of $v_{i}$ in the blue and red samples


## Ranking idea

Top and bottom 20 words according to

$$
p\left(\boldsymbol{v}_{i}\right)-p\left(\boldsymbol{v}_{i}\right)
$$

## Ranking idea

Top and bottom 20 words according to

$$
p\left(\boldsymbol{v}_{i}\right)-p\left(\boldsymbol{v}_{i}\right)
$$

to
women
right
senat
their
amend
woman
her
my
and
decis
famili
doctor
make
health
for
will
friend
court
law

## Ranking idea

Top and bottom 20 words according to

$$
p\left(\boldsymbol{v}_{i}\right)-p\left(\boldsymbol{v}_{i}\right)
$$



## Ranking idea

Top and bottom 20 words according to

$$
p\left(\boldsymbol{v}_{i}\right)-p\left(\boldsymbol{v}_{i}\right)
$$

## Ranking idea

Top and bottom 20 words according to

$$
p\left(v_{i}\right)-p\left(v_{i}\right)
$$

## Aside: "stopword removal" not recommended

- Very-frequent terms have been proving "increasingly" useful, e.g., for stylistic or psychological cues
- "a" vs "the" is surprising

Google Books Ngram Viewer

[for years LL assumed this was a bug, but see Language Log, Jan 3 2016:

## $\boldsymbol{p}\left(\boldsymbol{v}_{\boldsymbol{i}}\right)$ vs. count

```
to
women
right
senat
their
amend
woman
```

kill<br>not<br>procedur babi<br>of<br>abort<br>the

## $\boldsymbol{p}\left(\boldsymbol{v}_{\boldsymbol{i}}\right)$ vs. count

$p\left(\boldsymbol{v}_{i}\right)-p\left(v_{i}\right)$ favors big counts, i.e., $v_{i}$ towards the righthand side of this plot


## $\boldsymbol{p}\left(\boldsymbol{v}_{\boldsymbol{i}}\right)$ vs. count

$p\left(\boldsymbol{v}_{i}\right)-p\left(v_{i}\right)$ favors big counts, i.e., $v_{i}$ towards the righthand side of this plot
(can't have a large difference between two small differences)

Ranking by log odds-ratio

$$
\log \frac{p\left(v_{i}\right) /\left(1-p\left(v_{i}\right)\right)}{p\left(v_{i}\right) /\left(1-p\left(v_{i}\right)\right)}
$$

## Ranking by log odds-ratio



## Ranking by log odds-ratio



```
tonight
necessarili
martin
peter
leg
harvest
frist
bright
anim
trade
taught
dayton
obvious
4 0
industri
chines
admit
infant
```


(Move to handout: model choices)

## Aside: warning on ignoring (language) history

Should we really write $P\left(v_{i}\right)$, with no conditioning on context?

- Previous lectures: language accommodation/coordination
- Church 2000: "Empirical Estimates of Adaptation: The chance of Two Noriegas is closer to $\mathrm{p} / 2$ than $\mathrm{p}^{2}$ ". COLING.
- "Finding a rare word like Noriega in a document is like lightning. We might not expect lightning to strike twice, but it happens all the time, especially for good keywords."

Ranking by z-score of log odds-ratio, with model of variance (uninformative prior)

## Ranking by z-score of log odds-ratio,

 with model of variance (uninformative prior)| Women | of <br> dr <br> right |
| :--- | :--- |
| not |  |
| woman | partial |
| their | fact |
| decis | birth |
| famili | head |
| amend | you |
| her | perform |
| senat | born |
| friend | the |
| my | mother |
| choos | child |
| doctor | abort |
| durbin | kill |
| serv |  |
| pennsylvania | procedur |
| santorum | babi |

Ranking by z-score of log odds-ratio, with model of variance (uninformative prior)

| Women | of |
| :--- | :--- |
| dr |  |
| right |  |$\quad$ not | partial |
| :--- |
| woman |


Ranking by z-score of log odds-ratio, with model of variance (informative prior)

| women | aliv <br> deliv |
| :--- | :--- |
| woman | dr |
| head |  |
| right | perform |
| decis | head |
| her | perform |
| doctor | birth |
| durbin | healthi |
| choos | partial |
| santorum | child |
| v | born |
| pennsylvania | mother |
| pregnanc | abort |
| viabil | procedur |
| friend | pilival |
| privir | babi |
| taili | babi |



