

# Bias and Methods of AI technology studying Political Science

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# AI in political science

- **Political Science questions:**

- Coalitions forming/breaking up.
- Upcoming policy changes.
- Emerging political issues.
- Voter behaviour.

- **“Text-as-data”:**

- Twitter posts
- Transcripts of speeches
- Written parliamentary motions
- Surveys

- Data source and choice of method(s)?
- How does choice of ML method affect results?
- Differences if we study English language countries vs. smaller languages?



## Kollektivavtal

Enligt lagen 22/6 1928 avtal mellan arbetsgivarpart och fackförening eller annan liknande förening av arbetare om villkor som skola lända till efter rättelse för anställning av arbetare eller om förhållanden i övrigt mellan parterna. [Med arbetare förstås ej blott kroppsarbetare utan arbetstagare över huvud. Part på arbetsgivar sidan kan vara även stat, kommun eller annan menighet]

# Word embeddings from different parties

Exploratory pilot study presented at PolMeth Europe March 2021

- Learn word meaning as numeric vector - a word embedding.
- Similar meaning/use to give similar embeddings.
- How does embeddings differ when **trained on data from different parties?**



# Some Related Work

- **Rodman (2020):** *A Timely Intervention: Tracking the Changing Meanings of Political Concepts with Word Vectors*
  - word2vec embeddings to study meaning over time, ideological bias.
  - E.g. “Equality” in news article headlines 1855-2016.
  - Comparison to topic modelling - broad themes vs word meaning.
- **Rheault & Cochrane:** *Embeddings for Political Analysis*
  - Parliamentary speeches (US, Canada, UK).
  - “Party embeddings” for each party and session of parliament.

# Sweden and the US

- Sweden:
  - Proportional representation, coalitions.
  - Left - Right
  - Written motions from parliament.
- US:
  - Presidential and individual candidate elections.
  - Liberal - Conservative
  - Transcripts of Senate speeches.



# Data

- **US:**
  - Parliamentary debates (2015-2017)
  - Republicans: 19K docs, 8.5M tokens
  - Democrats: 17K docs, 8.4M tokens
- **Sweden:**
  - Written motions (2013-2017)
  - Moderates: 6K docs, 3M tokens
  - Social Democrats: 4K docs, 2M tokens.

# Methods

- Count-based:
  - PMI: co-occurrence matrix normalised by number of occurrences.
  - SVD: compressed PMI matrix (reduce noise)
- Neural Networks (word2vec):
  - CBOW: Learn embedding by training to predict missing word in given context.
  - Pre-training + fine tuning

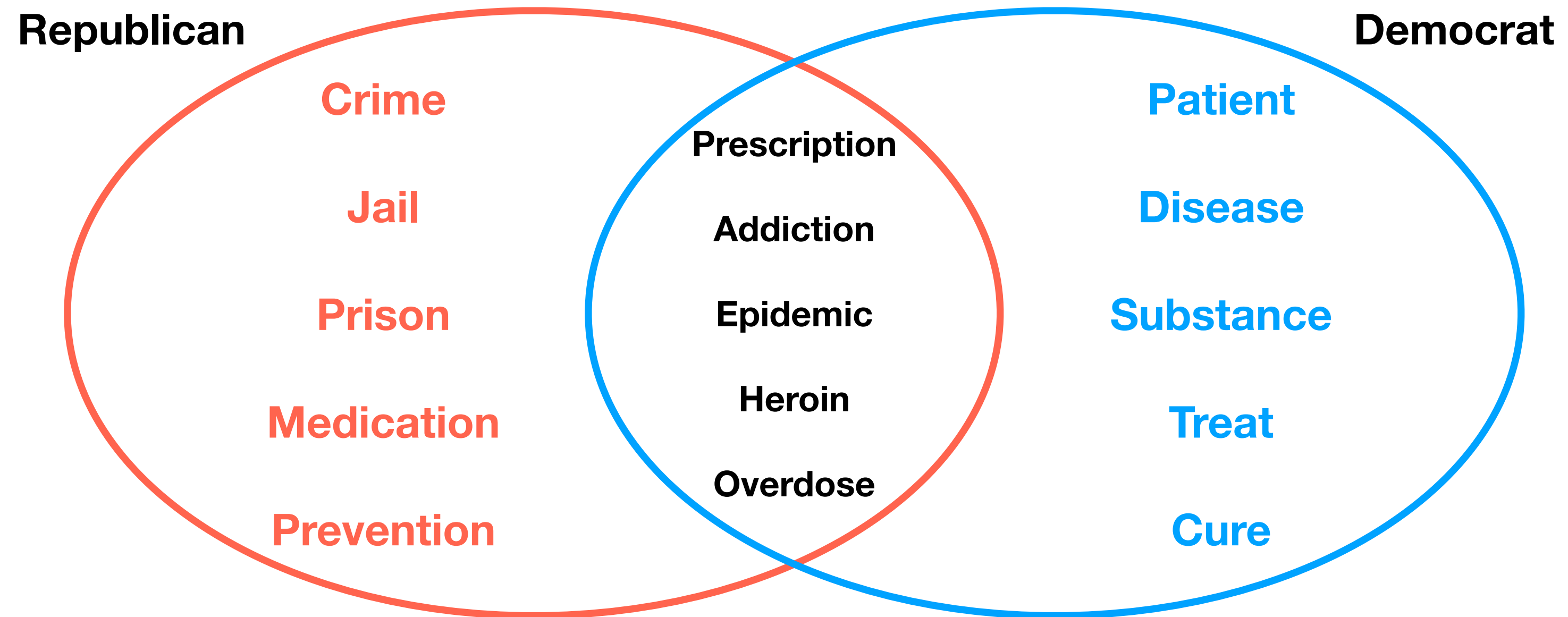


# Topics: Taxes and Drugs

- **Taxes:**
  - Expect division along left-right dimension.
- **Drugs:**
  - Not clearly a left-right issue.
- Measure cosine distance to these seed-words - find closest. Do they differ depending on which party used for training?



# Sample results: Drugs (US)



# Sample result: Taxes (Sweden)

**Social Dem.**

**Beskatta**  
*(Taxate)*  
**Avdrag**  
*(Deduction)*  
**Pensionär**  
*(Pensioner)*  
**Miljömål**  
*(Environmental targets)*  
**Nytta**  
*(Benefit)*

**Sänka**  
*(Reduce)*  
**Betala**  
*(Pay)*  
**Höja**  
*(Raise)*  
**Jobb**  
*(Job)*  
**Inkomst**  
*(Income)*

**Moderates**

**Skatteintäkt**  
*(Tax earnings)*  
**Löntagare**  
*(Wage worker)*  
**Konkurrens**  
*(Competition)*  
**Ekonomi**  
*(Economy)*  
**Finanspolitisk**  
*(Fiscal policy)*

# Summary

- Drugs topic similar pattern in US and Sweden.
- Taxes dimension more pronounced in Sweden.
- PMI model gave more interesting results detecting nuances in agreement/disagreement in word meaning.
- Neural Network tended to give example words (eg. listing different kinds of drugs).
  - Due to small window size, while PMI summarise whole document.

# Next Steps

## Data Science/ML: Explainable ML

- Political Science: mainly un-supervised problems.
  - How do we know our ML models are any good?
- Re-cast as **supervised learning**: learn which party wrote motion.
  - Test also large NN models, transformers, BERT.
- **Explainability techniques**: NN as black box, explain what characterise each party?
- Focus on **Swedish parliamentary motions**.
- Dealing with classifications of long documents - hierarchical NN's?

# Next Steps

## Political Science

- Analysis of “political influencers” in social media.
- First step: Brexit - qualitative analysis of posts by influencers, populists, parliamentarians.
- Social network analysis, polarisation, filter bubbles.
- Could serve as gold-standard for later quantitative analysis.



**Thanks for listening!**

Psst!  
Post-doc position  
will be announced  
soon!

