## TECHNICAL UNIVERSITY MUNICH

TUM Data Innovation Lab

## A Network Analytical take on the European Parliament

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## Motivation

Find Hidden Agendas

## Motivation



## Motivation



## Hidden Agenda

Following goal in non-obvious manner

## Hidden Coalition

Collaboration not appparent by direct work



## Gathering Data

|  | date | speechnr | agenda | name | nationality | party | euparty | text |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 1999-07-21 | en.19990721.1.3-001 | Address by the President | Nicole Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | Ladies and gentlemen; once again; I should lik... |
| 1 | 1999-07-21 | en.19990721.1.3-003 | Address by the President | Nicole Fontaine | France | Union pour la democratie française | Group of the European People's Party (Christia... | I thank the President-in-Office of the Council. |
| 2 | 1999-07-21 | en.19990721.1.3-005 | Address by the President | Nicole Fontaine | France | Union pour la démocratie françalse | Group of the European People's Party (Christla... | I am truly grateful; Mr Commissioner Marín. |
| 3 | 1999-07-21 | en.19990721.2.3-006 | Approval of the Minutes | Nicole Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | The Minutes of the last silting have been dist... |
| 4 | 1999-07-21 | en.19990721.2.3-007 | Approval of the Minutes | Marie-Hélène Gillig | France | Parti socialiste | Group of the Party of European Socialists | (FR) Madam President; with regard to the Minut... |

## http://linkedpolitics.ops.few.vu.nl

## Preprocessing

|  | date | speechnr | agenda | name | nationality | party | euparty | text |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 1999-07-21 | en.19990721.1.3-001 | Address by the President | Nicole <br> Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | [address, convey, heartfelt, trust, show, elec... |
| 1 | 1999-07-21 | en.19990721.1.3-003 | Address by the President | Nicole Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | [address, presidentinoffic] |
| 2 | 1999-07-21 | en.19990721.1.3-005 | Address by the President | Nicole <br> Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | [address, trull, grate, marn] |
| 3 | 1999-07-21 | en.19990721.2.3-006 | Approval of the Minutes | Nicole <br> Fontaine | France | Union pour la démocratie française | Group of the European People's Party (Christia... | [approv, last, distribut, comment |
| 4 | 1999-07-21 | en.19990721.2.3-007 | Approval of the Minutes | Marie-Hélène Glllig | France | Parti socialiste | Group of the Party of European Socialists | [approv, regard, yesterday, provid, inform, re... |



## Topic Modelling

- Latent Dirichlet allocation (LDA)
- Idea: Find topics in texts by assigning word probabilities to topics


Figure:
https://upload.wikimedia.org/wikipedia/commons/4/4d/Smoothed_LDA.png

## Optimal Number of Topics



## Topic Visualisation



Top-30 Most Relevant Terms for Topic 14 (1.8\% of tokens)


## Example



Figure: Danielle Auroi

Madam President, President
Prodi, ladies and gentlemen, after hearing Mr Prodi's proposals, I am utterly astounded by the position of the PPE and the PSE on food safety. Perhaps they do not feel capable of putting forward concrete proposals today, but we do. That is why we wished to propose a resolution for, throughout Europe, the series of scandals which have occurred means that, today, the citizens and consumers no longer have any confidence in their farmers. The quibbling involved in stating that the Committee ...

## Inferred Topics

## LDA result

(0, '0.475*"strategi" $+0.152^{*}$ "lisbon" $+0.055^{*}$ "object" $+0.030 *$ "implement" $+0.029 *$ "competit"')
(1, '0.153*"indian" $+0.130 *$ "threeyear" $+0.094 *$ "empir" $+0.066 *$ "disintegr" $+0.053^{*}$ "overshadow" $)$
(2, '0.028*"develop" + 0.020*"econom" + 0.018*"area" + 0.017*"support" + 0.011*"increas"')

(4, '0.307*"polish" $+0.239 *$ "domest" $+0.093^{*}$ "beekeep" $+0.091^{*}$ "gross" $+0.069 *$ "default"')
(5, '0.503*"medium" $+0.085 *$ "televis" $+0.069 *$ "broadcast" $+0.040 *$ audiovisu" $+0.036^{*}$ "guinea" )
( $6, ~ ' 0.202^{*}$ "marginalis" $+0.154^{* " w o r s e n " ~}+0.105^{* " a n t i d i s c r i m i n " ~}+0.066 *$ "perpetu" $+0.062^{* " m i c h e l " ') ~}$

(8, '0.161*"hamper" + 0.148*"smallscal" + 0.091*"anticorrupt" + 0.083*"bolster" + 0.077*"adr"')


## Inference

name date topic
0 Marie-Noëlle Lienemann 1999-07-01 [(23, 0.05672398), (38, 0.016829032$),(73,0.0 \ldots$
1 Gerhard Schmid 1999-07-01 [(141, 0.07714286), (242, 0.5914885), (257, 0....
2 Hanja Maij-Weggen 1999-07-01 [(36, 0.019324558), (109, 0.020725463), (111, ...
3 Ingo Friedrich 1999-07-01 [(60, 0.08798485), (110, 0.022743504), (144, 0...
4 Hans-Peter Martin 1999-07-01 [(146, 0.28848597), (238, 0.023322258), (242, $\ldots$


## Network modelling

| name | date | topic |
| ---: | :--- | :--- |
| A | $1999-07-01$ | $[(0,0.7),(1,0.3)]$ |
| B | $1999-07-01$ | $[(0,0.7),(1,0.3)]$ |
| C | $1999-07-01$ | $[(1,0.7),(2,0.3)]$ |
| D | $1999-07-01$ | $[(2,1.0)]$ |
| A | $1999-08-01$ | $[(0,0.5),(1,0.5)]$ |
| B | $1999-08-01$ | $[(0,0.3),(1,0.2),(2,0.5)]$ |
| C | $1999-08-01$ | $[(1,0.5),(2,0.5)]$ |
| A | $1999-09-01$ | $[(0,0.8),(1,0.2)]$ |

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| $\left[\begin{array}{c}0 \\ 1.2 \\ 0.8\end{array}\right]$ C <br> $\left[\begin{array}{c}2.0 \\ 1.0 \\ 0\end{array}\right] \mathrm{A}$ |  |  |
|  |  |  |

## Network modelling



## Community Detection



## Community Detection

$$
Q=\frac{1}{2 m} \sum_{v, u \in V}\left(a_{v u}-\frac{k_{v} k_{u}}{2 m}\right) \delta\left(c_{v}, c_{u}\right)
$$



## Outlier Detection



## Outlier

- Girvan-Newman algorithm
- High topic overlap $\Leftrightarrow$ Large edge weight
- Missmatching data


## Topic Distribution



## Topic Distribution



## Topic Distribution



## Outlier vs. Community



## Outlier vs. Community



## Outlier vs. Community



## Outlier vs. Neighbours



## Outlier vs. Neighbours



## Outlier vs. Neighbours



## Outlier vs. Neighbours



## Outlier vs. Neighbours



## Results Hidden Agenda



## Results Hidden Agenda



## Results Hidden Agenda



## Hidden Community Detection



Figure: From K.He et.al, Hidden Community Detection in Social Networks, 2017

## HiCoDe Algorithm

- Apply the base algorithm - Louvain Algorithm
- Calculate the modularity
- Weaken the structure by using refinement algorithms such as remove edge or reduce edge
- Repeat until appropriate layers


## Number of Layers

- Calculate the modularity for dominant community $Q_{0}$
- Perform T iterations of refinement and calculate modularity for each iteration $Q_{T}$
- Calculate average improvement ratio of modularity per iteration. as $R_{T}=\frac{\sum_{t=1}^{T} Q_{T}}{Q_{0} T}$
- Choose layer which has highest $R_{T}$


## Hidden Community Detection



## One Hidden Community



## Results Hidden Coalition



## Results Hidden Coalition



## Outlook

- Translation
- Metadata
- Hollistic Community Outlier


## Thank you and Questions?

