

Role Detection in Communication Networks

About specialising protestants and
deceived & deceiving employees

Ramona Roller
10th January 2023



Formal roles guide human behaviour

Enron



1

Kenneth Lay (**chairman**); Jeffrey Skilling (**CEO**)

Reformation



2

16th century **scholar**: Johannes Reuchlin

Image sources:

¹ [chron.com](https://www.chron.com): Enron's cast of characters

² [pforzheim.de](https://www.pforzheim.de): Johannes Reuchlin: Der Humanist aus Pforzheim

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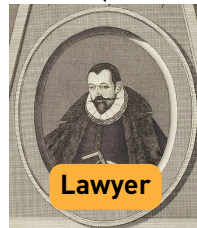
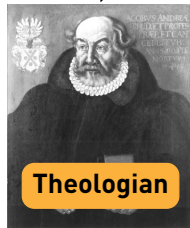
But what if humans deviate from their formal roles?

Enron: Explain deceit



1

Reformation: Explain differentiation



2

➔ Which other roles could explain this deviant behaviour?

Image sources:

¹ twitter.com/BayouCityHistory; [chron.com](https://www.chron.com)

² [wikipedia.com](https://www.wikipedia.com): Jakob Andreae; Heinrich Camerarius

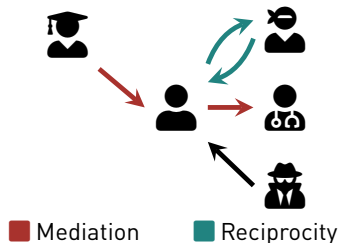
Informal roles in communication networks

▶ Informal role

- ▶ Emerge through **interactions** and are **context-specific**.

▶ Application

- ▶ **Communication network:** nodes = individuals; edges = message exchange
- ▶ Enron e-mail network; Reformation letter correspondence network
- ▶ **Communication role:** Group of nodes with similar topological embeddings

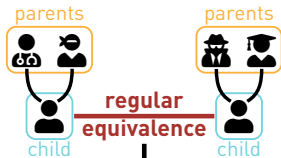


- ➔ How can we detect and interpret communication roles?
- ➔ How can we relate communication roles to observed behaviour that deviates from formal roles?

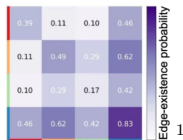
Role detection in networks: common practice (1/2)

Role detection: Procedure for grouping nodes in networks based on topological similarity

Structural approach



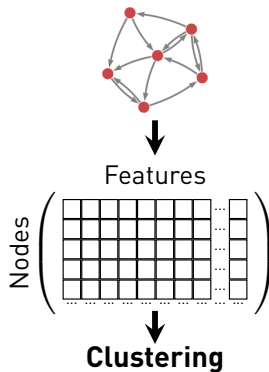
Stochastic Block Model



➔ Nodes are grouped into blocks (roles) but meaning of blocks is not clear.

➔ Open question: What do roles mean?

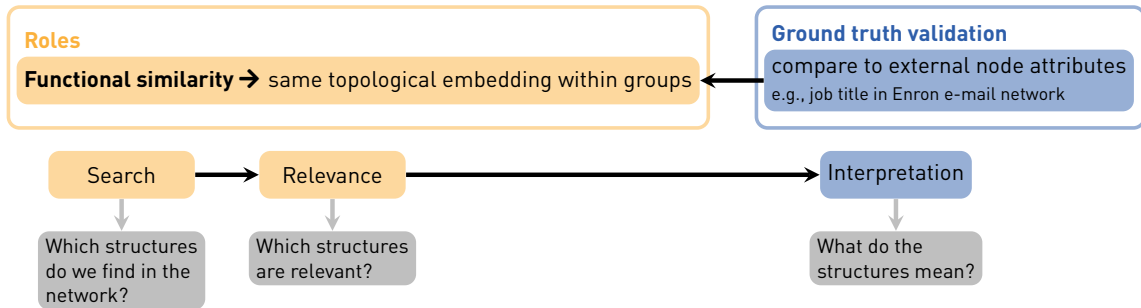
Feature-based approach



➔ Nodes are grouped into clusters (roles) but meaning of blocks remains technical.

¹ Image: Faskowitz et al., 2018: Weighted Stochastic Block Models ..., Nature.

Role detection in networks: common practice (2/2)



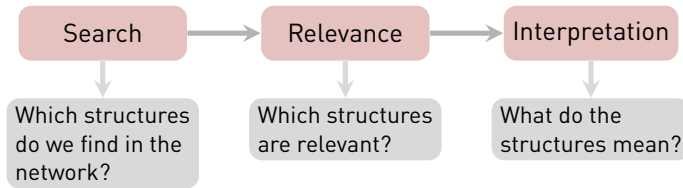
► **Problem:** Ground truth may not be relevant for detected roles¹.

➡ **Role detection is easy (1), interpretation is not (2).**

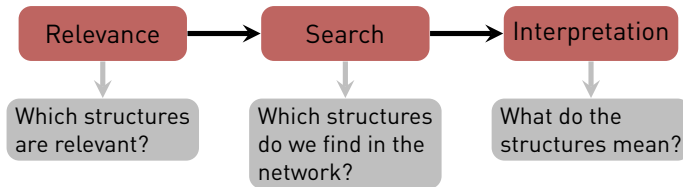
¹ Peel et al., 2017: The ground truth about metadata..., Science Advances.

Role detection in networks: New practice

Common practice



New practice



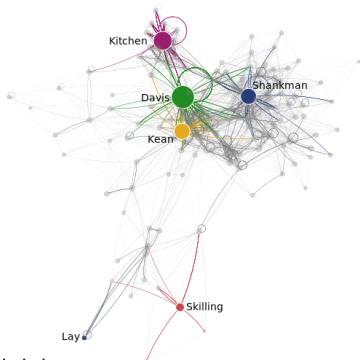
➔ **Select roles that are relevant in communication networks.**

Role detection in networks: applications

Enron e-mail network

- ▶ Enron: fraudulent commodities company
- ▶ Nodes: 144 employees
- ▶ Edges: 21,701 sent e-mails
- ▶ Period: 1998–2002

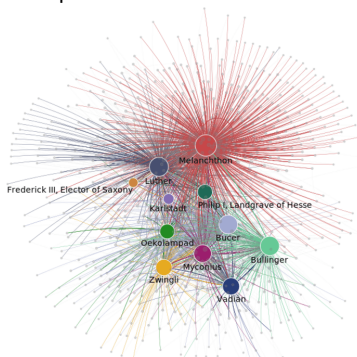
RQ Does deceit manifest in communication roles?



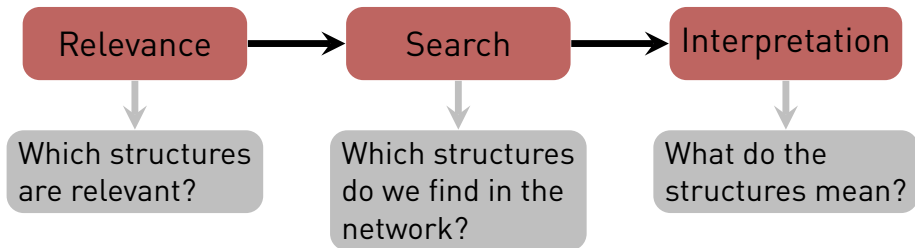
Reformation letter correspondence network

- ▶ Reformation: overthrow of Catholic Church
- ▶ Nodes: 3,000 scholars
- ▶ Edges: 30,000 sent letters
- ▶ Period: 1510–1565

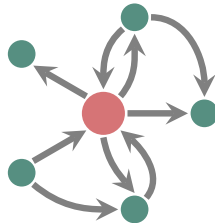
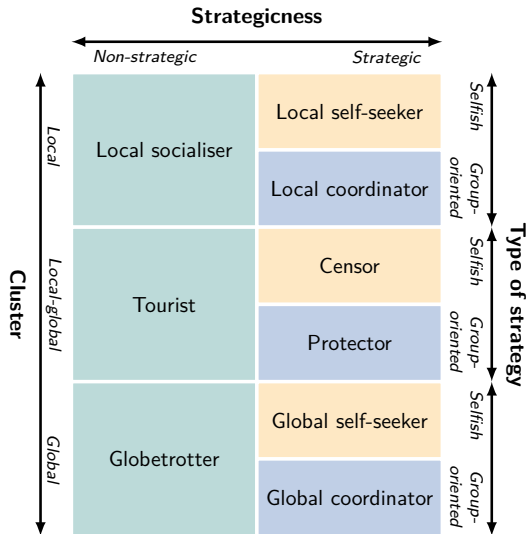
RQ Does differentiation manifest in the specialisation of communication roles?



Our pipeline



Relevance: Which roles are relevant?¹



Global coordinator

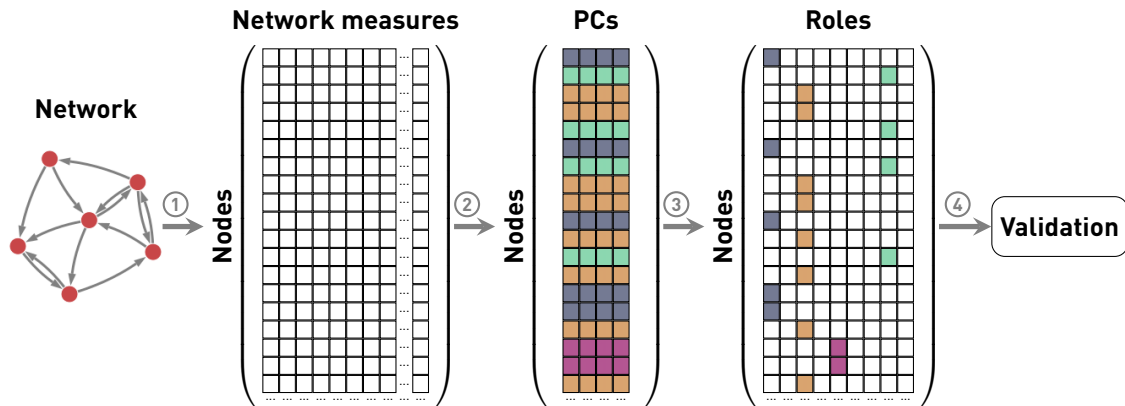
Mediates between local and distant parts of the network. Promotes information exchange between other nodes.

- ➡ Literature review of communication networks reveals to-be expected roles.
- ➡ Roles are relevant for communication networks and generalisable across them.

¹ Roller and Schweitzer, 2021: A new theory of social roles in networks, /.../socarxiv/xkafb/

Enron

Search: Which roles do we find in the network? (1/2)

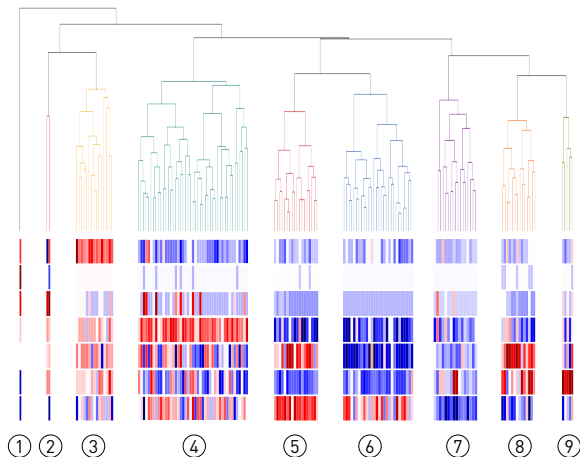


① Topological network measures for all nodes ② Dimensionality reduction (PCA) ③ Clustering

➔ **Roles: clusters of nodes based on non-redundant topological network measures**

Search: Which roles do we find in the network? (2/2)

- ▶ Result of clustering nodes along topological network measures
- ▶ **9 Columns:** Extracted roles
- ▶ **7 Rows:** topological measures in lower dimensional space (principal components (PCs))

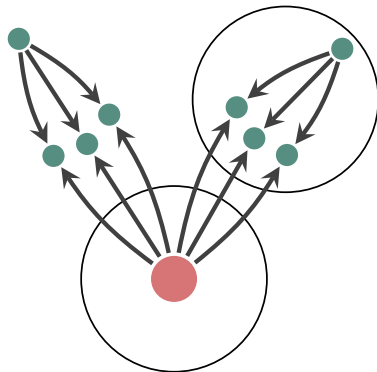


➡ **Nodes of the same role have similar values along the topological measures.**

Interpret topological measures in the lower dimensional space

Topological network measure	Factor loading
global binary out coreness	0.22
between binary out coreness	0.15
between mean in dice similarity	0.13
global mean out dice similarity*	0.83
between mean out dice similarity*	0.94
global out neighbours 1st order	0.17

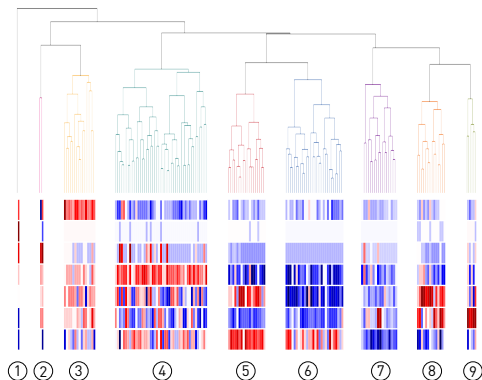
* Used for interpreting PC as communication constituent



- ➔ **Global action seeking:** Sending e-mails to those who also receive e-mails from others in the global network.

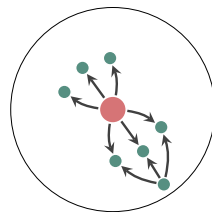
Interpret roles

- ▶ Per cluster identify pattern across topological measures in lower dimensional space (PC)
- ▶ Combine pattern with interpretation of PCs to yield a communicatio pattern → **role**



Role 5

- ▶ **PC 1:** - / - -
- ▶ **PC 2:** -
- ▶ **PC 3:** - -
- ▶ **PC 4:**
- - / - - -
- ▶ **PC 5:** + / + +
- ▶ **PC 6:**
- - / - - -
- ▶ **PC 7:** + + +



➔ **Role 5:** Contacts active communicators locally without receiving replies. ➔ **Local socialiser**

Roles of Enron employees



1



1

The bosses: Skilling & Lay

- ▶ Fraudulent behaviour, prison sentence
- ▶ Skilling: **local coordinator**; Lay: **ensor**

❓ **Too local?** Roles → lost awareness of leadership responsibilities?



2



3

Similar yet different employees: Kitchen & Shankman

- ▶ Kitchen: bonus + acquitted of charges; **global coordinator**
- ▶ Shankman: fired + prison sentence; **globetrotter**

❓ **Too unfocused?** Roles → presence/lack of clear tasks?



4

Loners: Kaminski

- ▶ voice of reason + not accused; **protector**

❓ Do roles mirror personal values?

➔ **No clear role pattern for deceitful behaviour: individual interpretations are necessary**

➔ **Missing information on behaviour of (crucial) employees**

e.g., Andrew Fastow (CFO) is not part of the data

¹ Skilling, Lay: [chron.com](https://www.chron.com)

² Kitchen: [bbc.com](https://www.bbc.com): Enron's internet monster

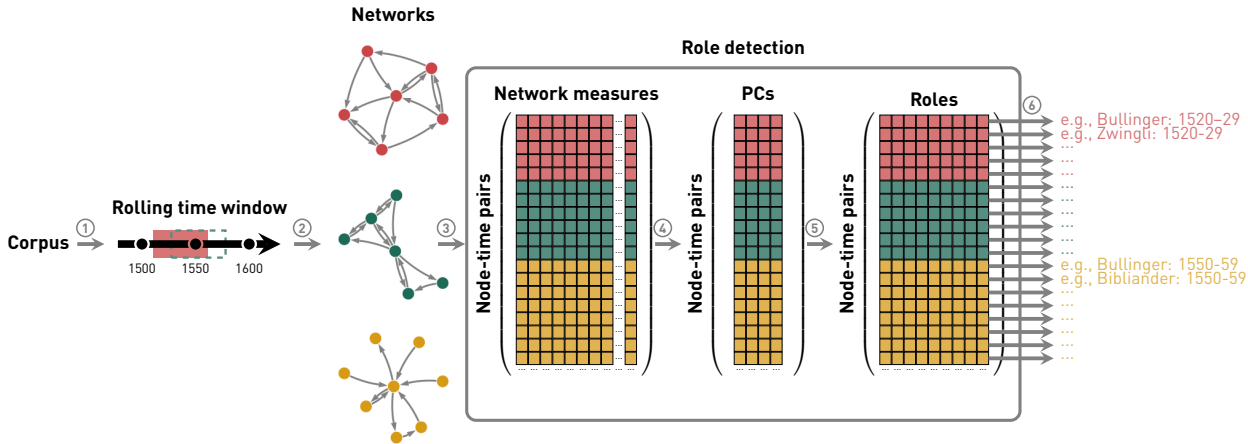
³ Shankman: [ladailypost.com](https://www.ladailypost.com): Ex-Enron Executive...

⁴ Kaminski: <https://business.rice.edu/>: Personal page

Reformation

Reformation: Role detection pipeline

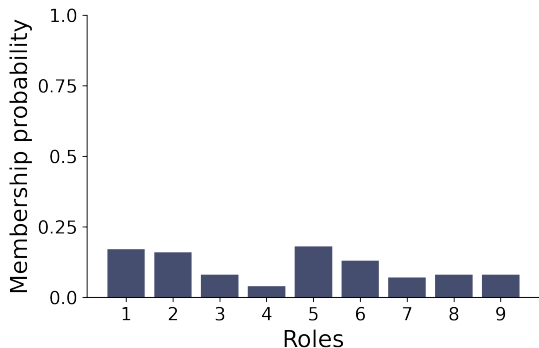
- ▶ Are communication roles subject to social differentiation?



- ➔ **Result of role detection:** Probability distribution across roles per reformer and subperiod.
- ➔ Probabilities indicate to what extent reformer fulfills roles.

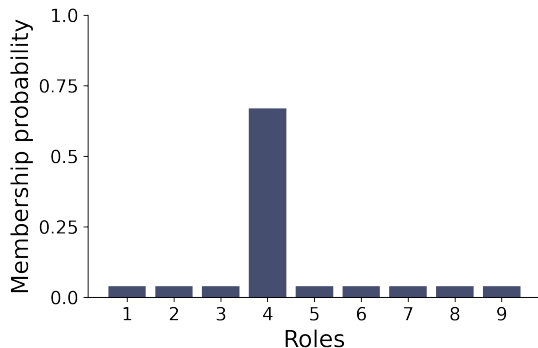
Role distributions of Heinrich Bullinger

1520–29: Uniform distribution



- ▶ Bullinger fulfills roles to *same* extent.
- ▶ Large entropy

1550–59: Nonuniform distribution



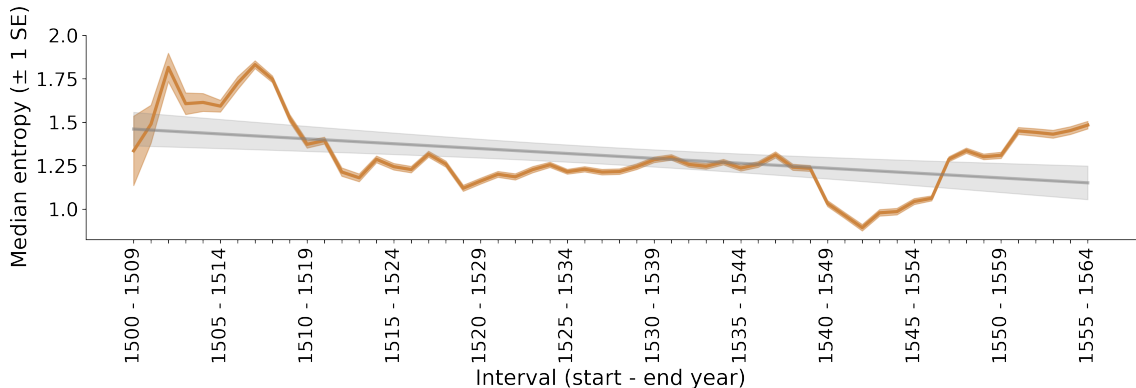
- ▶ Bullinger fulfills roles to *different* extents.
- ▶ Small entropy

➡ Entropy distinguishes between shapes of probability distributions.

Trend of entropy time series

Entropy time series

- ▶ Per subperiod: average entropies of probability distributions of reformers
- ▶ Result: one entropy value per subperiod

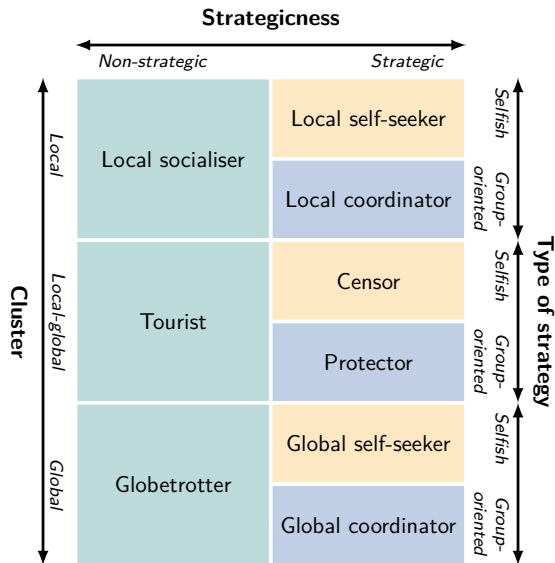


➡ **Negative trend:** reformers fulfill fewer roles over time → become more specialised in their communication roles

Conclusion

Our approach to role interpretation

- ▶ Roles are relevant for communication networks
- ▶ Roles are generalisable across communication networks
- ▶ **Enron:** local + non-strategic roles for fraudulent actors
- ▶ **Reformation:** individuals specialise in their roles → differentiation
- ▶ **Next step:** Theory building and falsification



MSc course: Data Science for Social Challenges

Simple inference

- ▶ OLS regression
- ▶ Gender pay gap
- ▶ Logistic regression
- ▶ Energy crisis
- ▶ Panel regression
- ▶ Loneliness

Causal inference

- ▶ Causal concepts
- ▶ Assessing the gain
- ▶ Diff-in-diffs
- ▶ Poverty
- ▶ Matching
- ▶ Political participation

Complex inference

- ▶ Network null models
- ▶ Human migration
- ▶ Measures and motifs
- ▶ Legalization of cannabis
- ▶ Network regressions
- ▶ Information overflow

For further information on the course (**syllabus, course catalog**) visit <https://www.sg.ethz.ch/teaching/data-science-for-social-challenges/>