



Complex networks in audit

A data-driven modelling approach

With you today

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Introduction

Agenda

- Introduction to auditing and financial information
- A network model for auditing
- Network statistics
- Network coarsegraining
- Network similarity



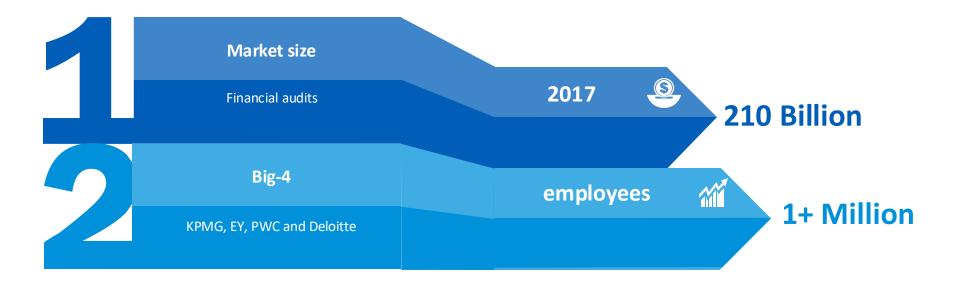
Introduction

Financial information

1 Lendors Investors **03** Creditors **04** Governments trust



Audit industry



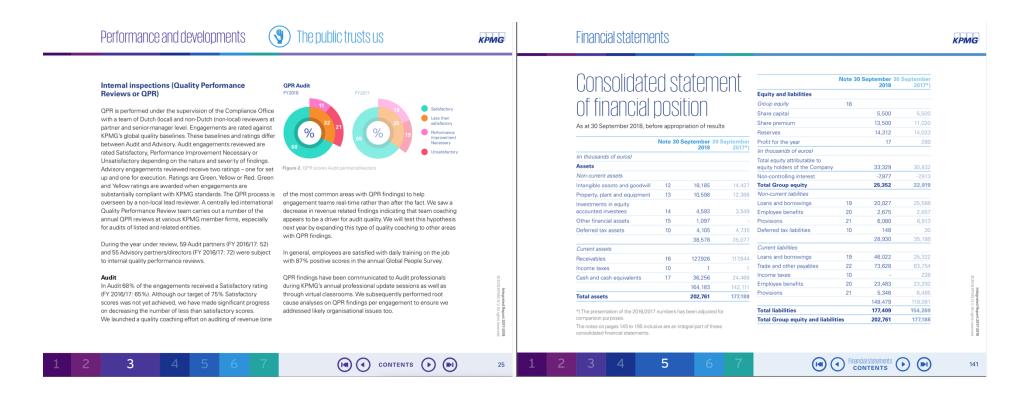
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- https://www.ey.com/en_gl/global-review/2018



What is an audit?

Annual statement

True and fair representation





What is an audit?

"Auditing is the accumulation and evaluation of evidence about information to determine and report on the degree of correspondence between the information and established criteria. Auditing should be done by a competent, independent person. - " [4]

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Installed new kitchen sink (hours)

Nest smart thermostat



ISA 520 describes the audit objective of the substantive analytical procedures as being:

 "To obtain relevant and reliable audit evidence when using substantive analytical procedures; and"

 "To design and perform analytical procedures near the end of the audit that assist the auditor when forming an overall conclusion as to whether the financial statements are consistent with the auditor's understanding of the entity." - ISA 520

"IFRS 15 establishes the principles that an entity applies when reporting information about the nature, amount, timing, and uncertainty of revenue and cash flows from a contract with a customer. Applying IFRS 15, an entity recognizes revenue to depict the transfer of promised

ects the consideration to which the r services.

e following five steps:

ract. Performance obligations are ods or services that are distinct.

Tank you for your business!

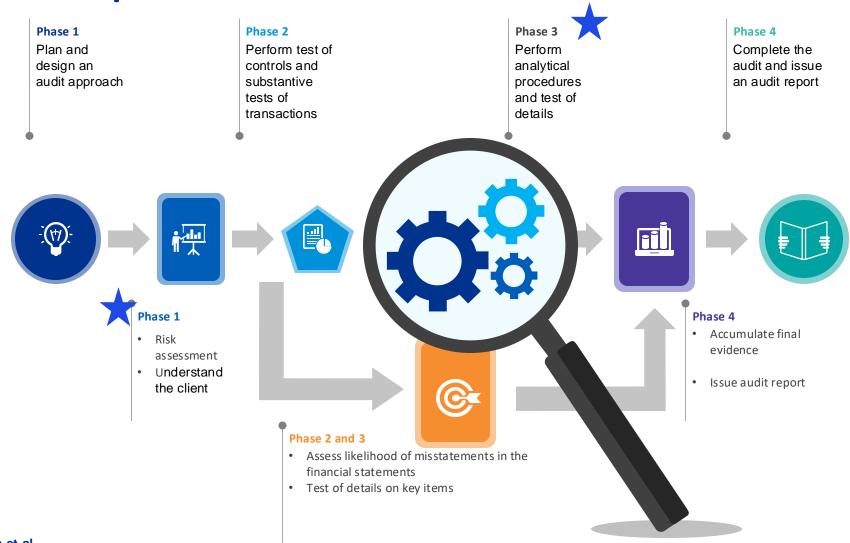
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Audit phases



Arens et al.



The need

A recent report states, among other things, that new technology can improve audit quality. Specifically, it indicates that data-driven methods and analysis could make audits more effective and efficient.

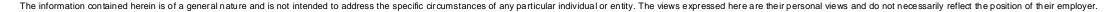


Bron: https://www.accountant.nl/nieuws/2018/6/reacties-op-mca-rapport-tijd-voor-fundamentele-verandering/



Limitations of current procedures

- Manual in nature (sampling of invoices, counting of inventory)
- Cannot always be directly digitized one-on-one
- They study the financial system as a set of independent components





Activities of a company

Revenue

invoices





A network model

Data sources of an audit

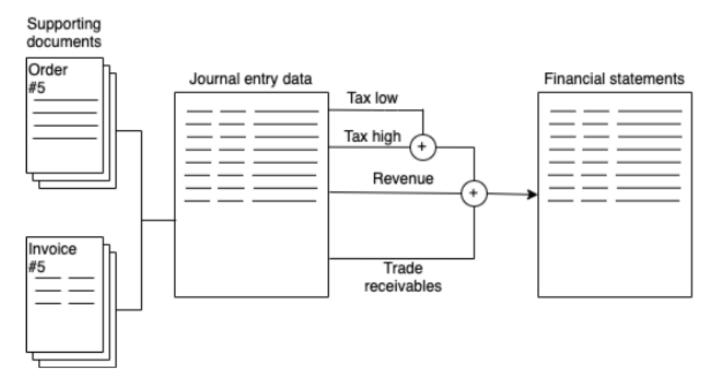


Figure 1.2: This high-level data overview shows how the aggregated information is connected to the documents recorded in the company's systems. The financial statements (right) are an aggregate representation of the journal entry data (middle), which is connected to supporting documents such as invoices, work orders, and more (left).



The data: Journal entries

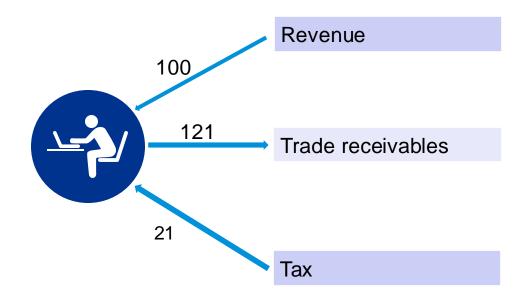


| | Financial Account | Journal | Date | Debit | Credit |
|--|-------------------|--------------|------------|-------|--------|
| | Revenue | Sales ledger | 02-11-2019 | - | 100 |
| | Trade receivables | Sales ledger | 02-11-2019 | 121 | - |
| | Tax | Sales ledger | 02-11-2019 | - | 21 |

Boersma, Marcel, et al. "Financial statement networks: an application of network theory in audit." The Journal of Network Theory in Finance 4 (2018).



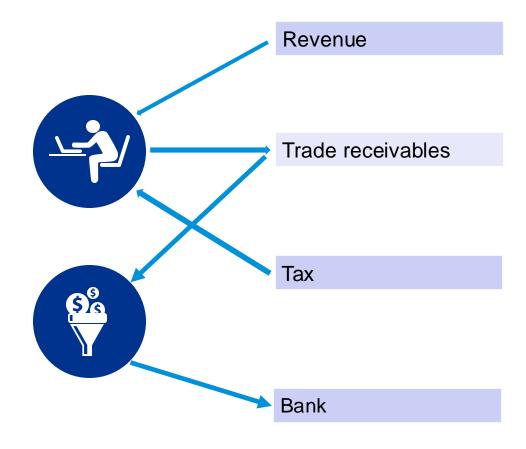
The data structure: a bipartite network



Boersma, Marcel, et al. "Financial statement networks: an application of network theory in audit." The Journal of Network Theory in Finance 4 (2018).



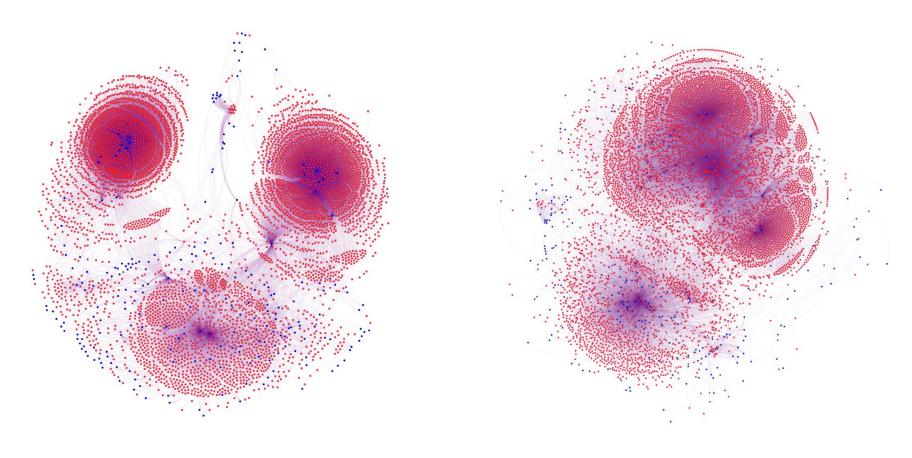
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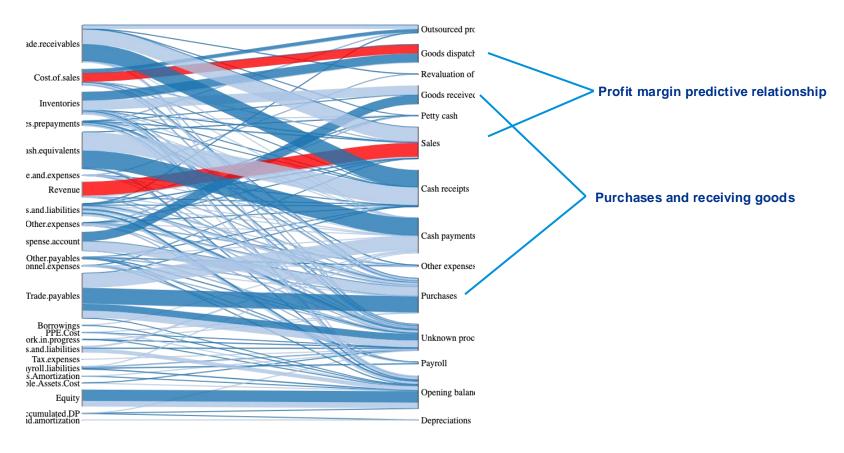
Real networks



Boersma, Marcel, et al. "Financial statement networks: an application of network theory in audit." The Journal of Network Theory in Finance 4 (2018).



Simplified network



Boersma, Marcel, et al. "Financial statement networks: an application of network theory in audit." The Journal of Network Theory in Finance 4 (2018).



Audit insights

Auditors can use the network for risk assessment by:

Comparing their understanding of a client's financial structure with the network structure

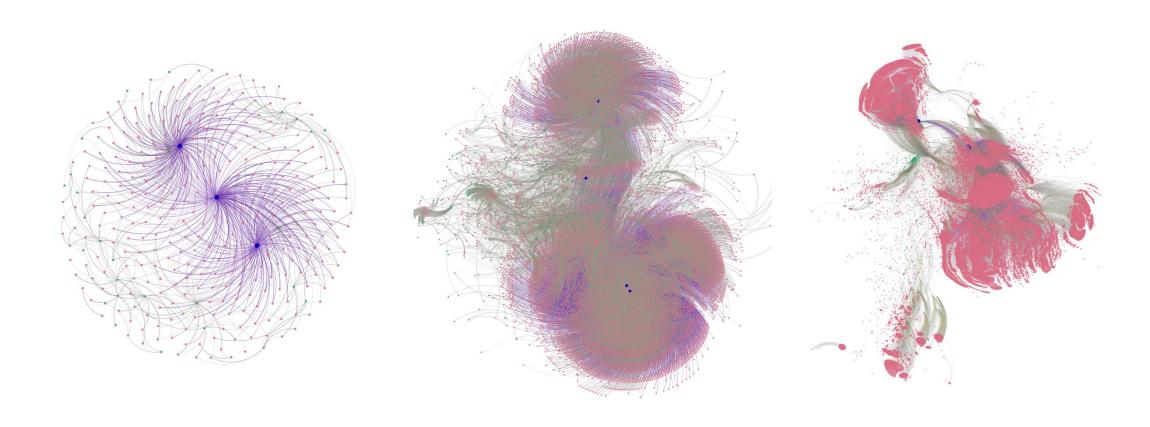
Auditors can use the network to obtain substantive evidence by:

A way to assess the concistency between financial flows for audit evidence



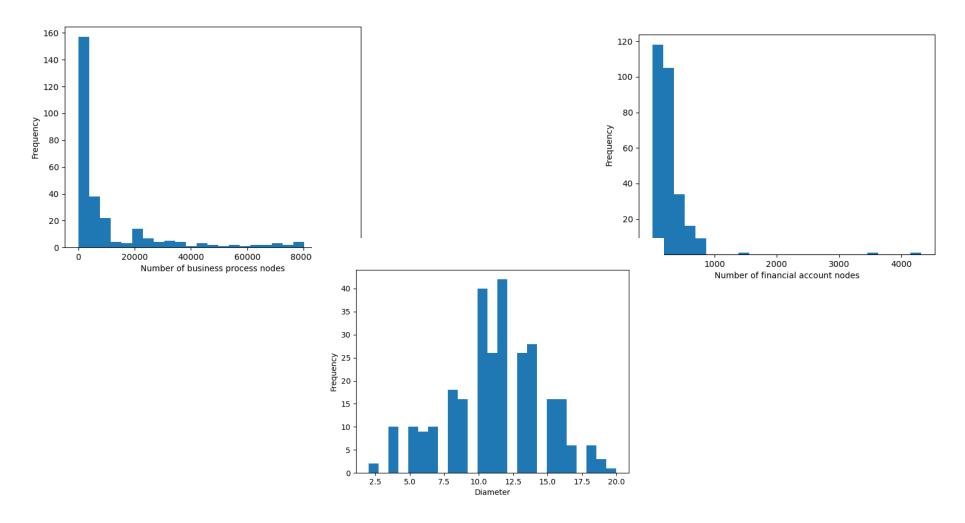
Network statistics

Understanding the networks





Baseline statistics





Baseline statistics: degree distribution

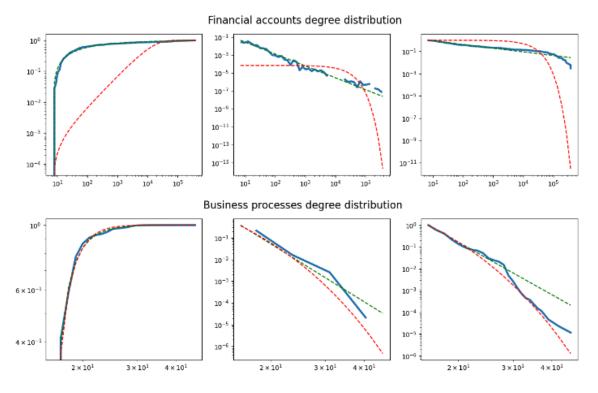
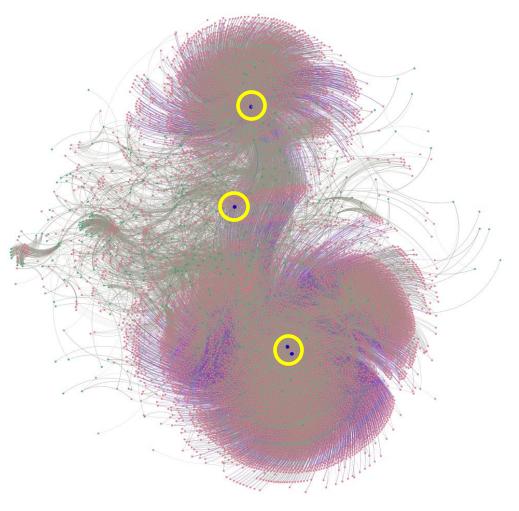


Figure 8. On the left is the cumulative density function, in the middle is the probability density function, and on the right is the complementary cumulative density function for financial statements network 62 (large) with 391.688 nodes in total. The top row is the financial account nodes distributions, and the bottom row is the business process nodes distributions. The green dashed line is a power-law distribution, the red dashed line is an exponential distribution, and the solid blue line is the empirical distribution. Note that the axes are on a log scale.



Analysis: what are important parts of the financial st



Financial

- hubs
- gatekeepers
- core-activities



Results and conclusion

- Number of business process nodes determines the size
- An increase in the number of nodes does not result in a corresponding proportional increase in the network's diameter
- we found evidence of a heavy-tailed degree distribution in financial account nodes, leading to hubs of interest to auditors
- Results confirm, for a small sample, that the centrality measures highlight important hubs in the financial structure, increasing the auditor's understanding of the company.



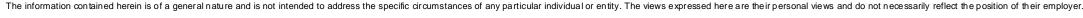
Audit insights

Baseline statistics:

• used to assess whether a new client's financial structure is in line with expected statistics (power-law distribution of financial account nodes, diameter expectations, etc).

Centrality measures

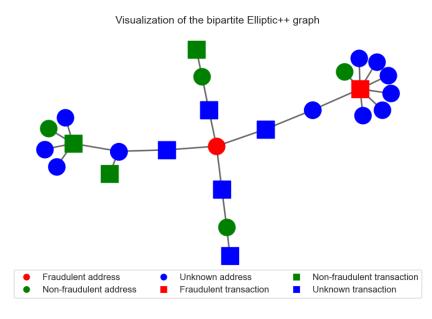
- Financial Gatekeepers (betweenness centrality)
- Financial Hubs (closeness centrality)
- Financial Core-activities (degree centrality)





Semi-supervised Anomaly Detection with Granger Causal Explanations for Financial Networks We propose a semi-supervised graph neural network (GNN)

We propose a semi-supervised graph neural network (GNN) that learns to flag fraudulent items based on limited labeled and unlabelled data points. The flagged items are explained by highlighting related items in the network using Granger causality. We use a real-world dataset, Elliptic++, based on Bitcoin transactions, to demonstrate the algorithm. We show that we obtain high-performance scores, but more importantly, we provide causal explanations for identified fraud based on graph structure. This capability can help auditors thoroughly investigate the implicated group and its interactions, offering a comprehensive understanding of the anomalies.



Nguyen et al. 2024 (work in progress)



Network coarse-graining

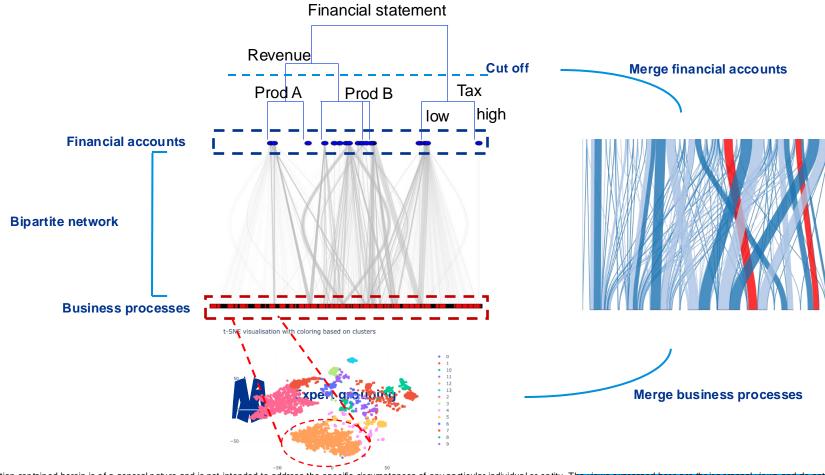
Network coarse graining

Motivation: Simplify the detailed networks to provide a high-level overview of the financial structure

Results: We found an effective way to coarse-grain the network of detailed financial statements. The method we proposed groups the nodes in the network in an automated way; the resulting groups show a high resemblance with the groups annotated by an auditor. We resolved the detailed financial statements network limitation, a manual coarse-graining procedure. In addition, the derived node embeddings can be hierarchically ordered, a valuable property in case the auditor wants to adjust the granularity of the high-level representation. The auditor can use this property to dynamically explore a client's financial structure.



Simplifying the network



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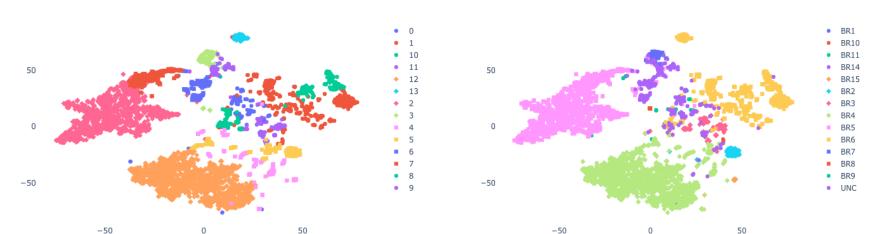
Expert vs clusters: company A

Clusters

Expert

V-score: 0.71

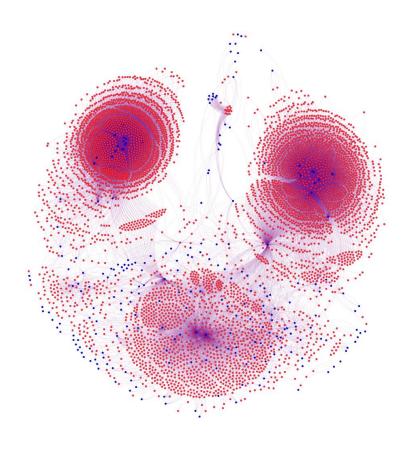
t-SNE visualisation with coloring based on clusters

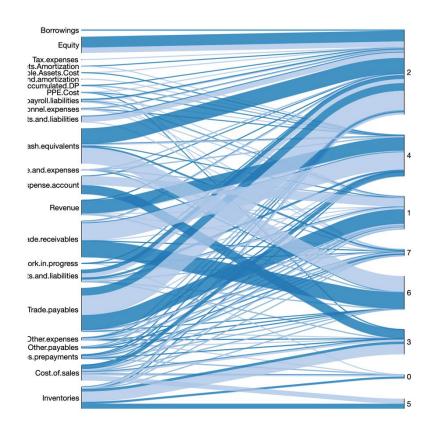


t-SNE visualisation with coloring based on expert labels



Results







Network similarity

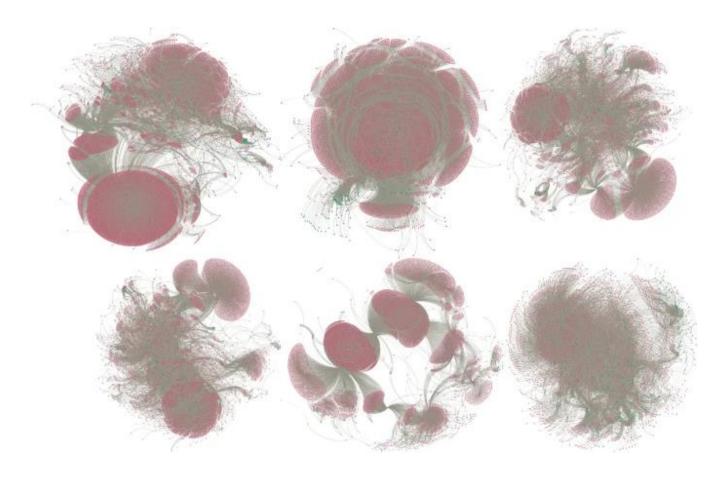
Network similarity

Motivation: Measure cross-sectoral structural similarities from financial networks

Results: Our approach is based on the analysis of 300+ real transaction datasets that provide auditors with relevant insights. We detect significant changes in bookkeeping structure and the similarity between clients. For various tasks, we obtain good classification accuracy. Moreover, closely related companies are near in the embedding space while different industries are further apart suggesting that the measure captures relevant aspects.



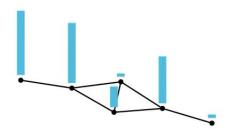
Are these networks similar?





From network to distribution

$$a^{h+1}(v) = rac{1}{2}igg(a^h(v) + rac{1}{deg(v)}\sum_{u\in N(v)}w((v,u))a^h(u)igg)$$

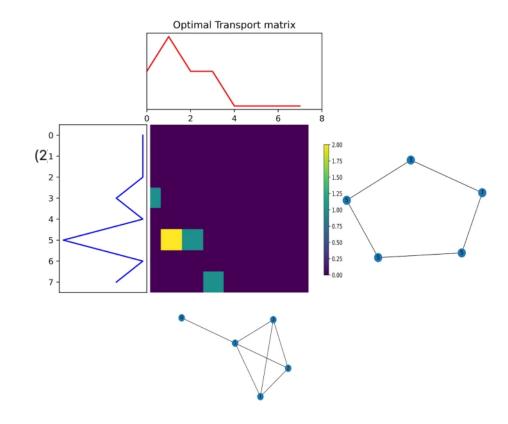




How do we measure similarity?

We measure the earth-movers distance.

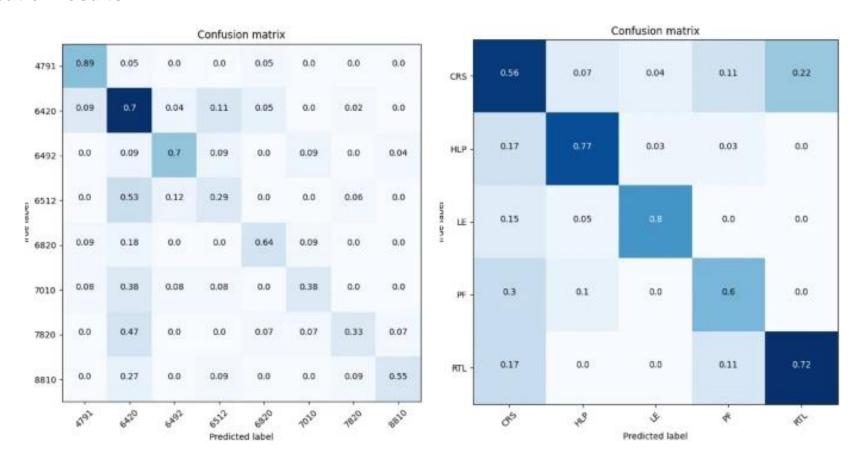
$$W_{\mathtt{p}}(\sigma,\mu) = \left(\inf_{\gamma \in \Gamma(\sigma,\mu)} \int d(x,y)^{\mathtt{p}} d\gamma(x,y)
ight)^{rac{1}{\mathtt{p}}}$$





The similarity measure results

Classification results:





Audit insights

Auditor can use this for assessing the audit risk by:

- Comparison with industry peers
- Comparison with prior year network structures



Financial statement networks

Outlook and discussion

Contributions

- A generic network representation of transaction data
- Base line statistics of Financial statements network
- A way to coarse-grain the network to provide a high-level understanding of the financial flows
- Similarities across companies

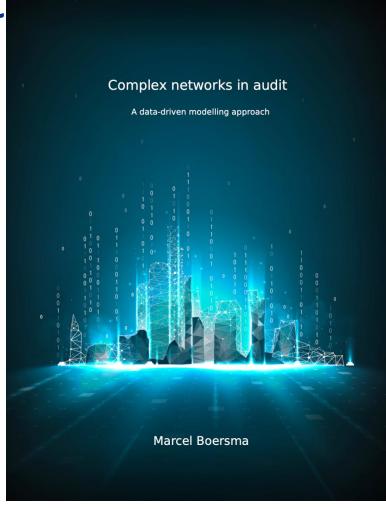
Conclusion: Our research focuses on developing data-driven audit methods to enhance the overall quality of audits. We achieved this by revealing the financial structure as a bipartite network. We answered multiple relevant audit questions by analyzing the network structure, demonstrating the chosen representation's usefulness. We showed that the financial statements network could be applied to risk assessment procedures.

Outlook:

- Financial statements network for fraud analysis in audit
- Assurance from models



Data driven audits — paving the way for higher qualit







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