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# Coordination in physician patient-sharing networks and the continuity and quality of care for type 2 diabetes patients:

A Social Network Approach



#### Content

- Background
  - Networks, cooperation, and coordination
  - Aim
- Methods
  - (Quantitative) Social Network Analysis
  - Data and network construction
  - (Network) measures
- (Preliminary) results
- Conclusion and discussion



## Background

#### **Coordinated care is necessary to deliver best care for diabetes** patients

- Within professions
  - Sharing knowledge / practices
  - Sharing patients
- Between professions
  - Sharing patients
  - Exchanging information of patients



## Background

- Social Network Analysis can provide insight in how physicians actually collaborate
  - With which other physicians?
  - Patterns of collaborating physicians
  - Who is more 'connected'?

#### How to collect data on networks?

- Method: patient-sharing networks
  - Based on healthcare claims
  - Connections formed by shared patients

#### However: how to interpret those networks?



### Background

- Meaning and use of patient-sharing networks
  - Defining collaboration groups
  - Indication of coordination

#### Sharing ≠ Coordination

- Referring as coordination
  - Characteristics of a strong primary care system
  - Enhance coordination

#### Data so far unsuitable



#### Background Research aims

- To explore variation in how physicians who treat type 2 diabetes patients form networks with other physicians, both intra and inter-disciplinary.
- To study how coordination of care in the form of referrals takes place within those networks.
- To test if networks in which physicians refer patients more often perform better?



# What is Social Network Analysis?

- Method with roots in the Social Science
- Underlying idea: People act not independently but are influenced by others
- These patterns of connections can be studied
  - Sociology, biology, computer science, neurology
- Quantitative method
  - Nodes  $\rightarrow$  Actors
  - Ties  $\rightarrow$  Connections
- Outcomes
  - Network parameters
  - Sociograms



# Methods Data

- Reimbursement data AOK BW health insurer
  - 4 million patients
- Included physicians:
  - General physicians
  - Internists
  - Ophthalmologists
- Creating physicians' patient-sharing network
  - Connection when sharing >4 patients



## Methods Networks levels

- Full network
  - One network including all physicians and links
- Ego network
  - Only including directly connected physicians to a specific node
- Local communities
  - Smaller subnetworks
  - Using a multi-level modularity optimization algorithm for finding community structures, sets of densely connected nodes were detected



### Methods *Measures: referrals*

**Identifying referral ties** 

- 1. Remove 'self-referrals'
- 2. Select only: GP  $\rightarrow$  Specialist ties
- 3. Timeframe < 3 months
- 4. Select only: >1 referral tie



# Methods

#### Network measures

- Ego network characteristics
  - Number of connections
  - Density, Betweenness, Constraint
  - Proportion of referred ties
- (Sub)Network characteristics
  - Density, Centrality
  - Proportion of referred ties



# Methods – Measures

- Physician characteristics:
  - Physician's specialization
  - Age and Gender
  - Involvement in a type 2 diabetes program
- Outcomes at patient level
  - Continuity of care: number of physician visits
    - number of different physicians per specialization
  - Suboptimal treatment Complications
    - Hospitalization



# (Preliminary) Results Descriptive

- Included 9260 physicians
  - 79% GP
  - 13% internists
  - 8% ophthalmologists
- Physicians' age 55
- 65% male
- Network connections
  - 1.874.333 shared patients
  - Forming 237.162 links between them
  - Average 51 connections



# (Preliminary) Results Structure and specialization



(Prelin Specia	ninary <i>lizatio</i>	) Resul	ts Struct	ure			
	GP (pink)	Intern. (orange)	Ophthal. (green)				
Number of connections	36.6	96.1	123.2				
Density	0.74	0.48	0.35				
Constraint	0.23	0.13	0.07				
Proportion referrals	0.76	0.57	0.60				
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# (Preliminary) Results Community structures

# (Preliminary) Results Community structures

	Mean	Min	Max	
Size	489	94	952	
N connections	50.8	38.5	63.2	
Density	0.13	0.04	0.40	
Centrality	0.50	0.32	0.96	
Prop. GPs	0.79	0.74	0.82	
Prop. referrals	0.72	0.62	0.79	
Prop. in DMP	0.64	0.54	0.71	





![](_page_18_Figure_0.jpeg)

# (Preliminary) Results Determinants of number of phycisians

![](_page_19_Figure_1.jpeg)

![](_page_19_Picture_2.jpeg)

# (Preliminary) Results Determinants of complications

![](_page_20_Figure_1.jpeg)

# (Preliminary) Results Determinants of hospitalization

![](_page_21_Figure_1.jpeg)

![](_page_21_Picture_2.jpeg)

### Conclusion

- Social Network Analysis can capture (variation in) physician collaboration networks
  - Between and within specializations
  - Between Community structures
- Referring
  - Can be distinguished from sharing patients
  - Varies between physicians and community structures
  - Can be linked to better coordination/continuity of care
  - Can be linked to better patient outcomes

![](_page_22_Picture_9.jpeg)

## Discussion

- Strengths and limitations
  - Networks based on shared patients
  - Only data from one health insurer
  - Few physician characteristics
- To discuss for future research
  - Theoretical link network measures and outcomes
  - Methodology: strength of connection, sharing and referring
  - Other purposes using Social Network Analysis?

![](_page_23_Picture_9.jpeg)

Discussion Take home

# Networks and coordination matter

![](_page_24_Picture_2.jpeg)

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