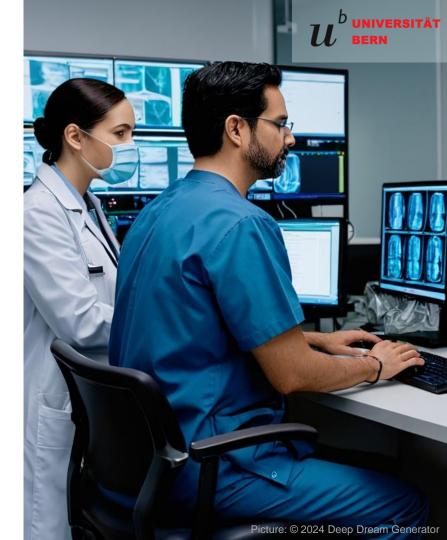
Unravelling the Human Interaction with Generative AI-Based Decision Support in Healthcare:

Types of Chat Users When Having ChaGPT vs. a Human Expert as a Chat Partner



#### **Mayra Spizzo**

CAS Applied Data Science, Module 5

# 5-15%



#### How can we reduce that number?

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- At the same time, AI systems are increasingly implemented in various industries
- Also in the field of medicine, Al-based systems are tested for different purposes
- How could AI be used to reduce the number of incorrect patient diagnoses?
- One approach could be the implementation of a LLM like ChatGPT
- This leads to the question: What do we know about ChatGPT and its implementation in the diagnostic decision-making process?

## $u^{b}$

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### How can we reduce that number?

- Accuracy of ChatGPT for differential diagnoses is 60.3%
- But at the same time, there is potential in augmenting diagnostic decisions
- Difference to traditional AI system: Involvement of the human in the process of the recommendation generation



How and why do the interactions with generative AI differ from the interactions with a human coach?

Ferdush et al. (2024), Rao et al. (2023)

 $u^{b}$ 

### Data collection

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- Online experiment
- With fourth-year medical students from the Charité Medical School in Berlin
- N = 158
- Diagnostic task: Solve 2 patient cases based on real emergency cases
- Conditions: ChatGPT vs. Human coach

### Method



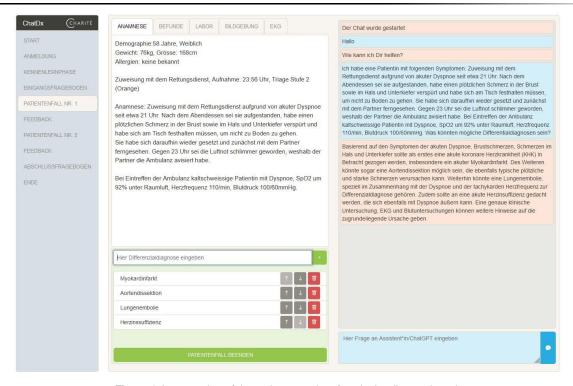


Figure 1 A screenshot of the patient case interface in the diagnostic task

### Method

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#### Patient case

Patient history
Blood samples
Laboratory results
Medical imaging
ECG

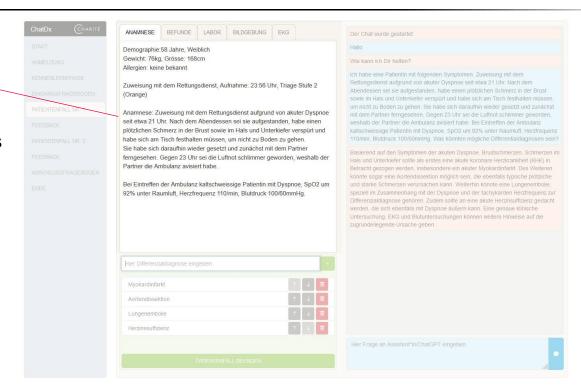


Figure 1 A screenshot of the patient case interface in the diagnostic task

BERN

### Method



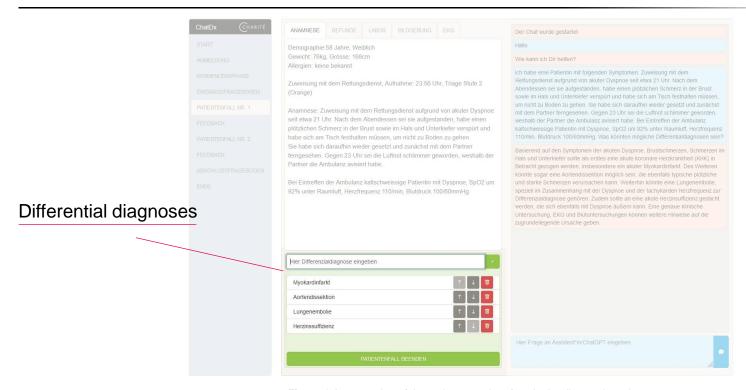
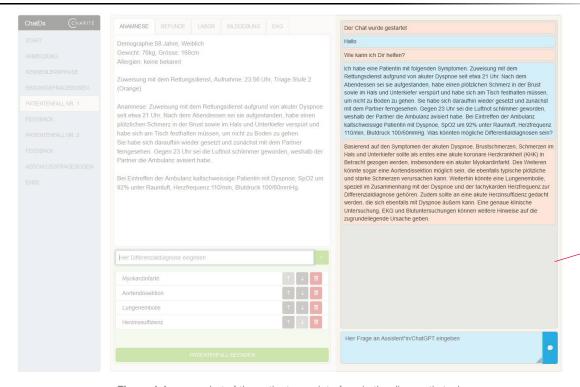


Figure 1 A screenshot of the patient case interface in the diagnostic task



### Method





#### Chat

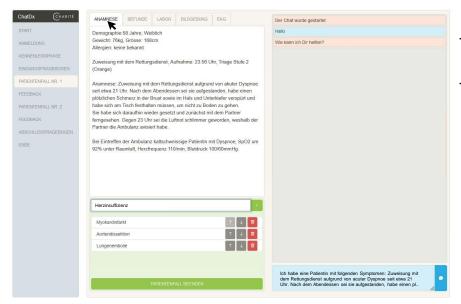
In real time with assigned assistant (ChatGPT or human)

Figure 1 A screenshot of the patient case interface in the diagnostic task



### Method





Time	Action	Additional information	
15:15.21	Click on patient record	Tab patient history	



### Data analysis

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- Chat interactions are coded with MAXQDA
- Coding scheme is used
- Intercoder agreement is calculated
- In addition, an algorithm is used to classify the chat interactions
- Further variables:
  - Condition
  - Patient information acquisition
  - Duration of chat interaction
  - Amount of differential diagnoses
  - Accuracy of diagnoses

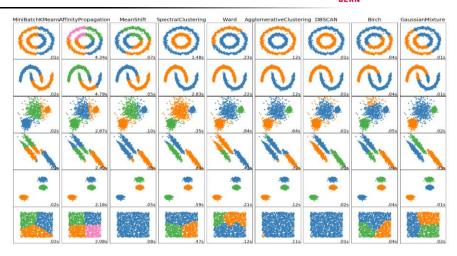
	DERING THE PROPERTY OF THE PRO
Category	Explanation
Technical questions: Close a knowledge gap	A general question without connection to a diagnosis
Request: Support for generating a differential diagnosis	A request to the assistant that he generates/suggests a new differential diagnosis
Statement: Expression of a suspected differential diagnosis	A statement by the user on its own or as an answer to a question from the assistant
Differentiation: Support for differentiating between differential diagnosis	A minimum of two differential diagnoses are mentioned - question about differences between the differential diagnoses
Verification: Support for checking and excluding differential diagnoses	Based on a diagnosis the question is asked about diagnostic findings (wanting a diagnosis to be verified out of the context)
Diagnostics: Support for selecting diagnostics	Based on a diagnosis the question is asked about further diagnostics
Management: Support for the management and the next steps being taken	A general question about further steps needed to being taken, or about special measurements after a certain diagnosis



### Data analysis

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- Clustering techniques
  - K-Means
  - Dendrogram
  - Further techniques according to PCA
- For each technique 3 clustering models
  - Overall model
  - ChatGPT-only
  - Human-only



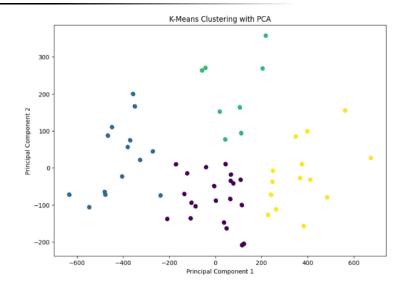
Seif (2018)



### First preliminary analysis

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- K-means with 114 data points
  - 4 Clusters
  - Differences between conditions observable
- More data is needed for validation and to support the assumptions







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Preregistration

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