

# Leveraging Large Language Models for Logistics Information Extraction: A Case Study on Two International Disasters

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# Background

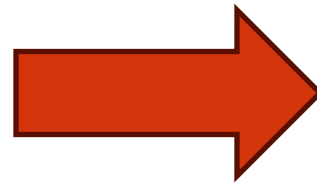
- Situational awareness during disasters requires quick synthesizing of information from multiple sources

- Information management officers' triage, interpret, and convert text into structured, mission-usable products.
- Can a LLM accomplish the same task?

“The airport in Bukavu is now fully operational”

“The Logistics Cluster coordinates the transshipment hub in Goma.”

Unstructured narrative reports from multiple sources



Common operating picture:  
Maps of logistics infrastructure

# Research Question

- **How effectively can ChatGPT extract relevant infrastructure and logistics information from text-based disaster reports?**
  - How do model configuration parameters, such as version and temperature, affect its performance?
  - What strategies can enhance the performance?

# Literature Review

- Studies highlight LLM capabilities in various tasks:
  - Geolocation extraction – Yin et al. (2023)
  - Temporal relationship identification – Yuan et al. (2023)
  - Dialogue generation – Bai et al. (2023)
  - Annotation – Gilardi et al. (2023); Labruna et al. (2023)
  - Decision-making in wargame simulations – Lamparth et al. (2024), etc.
- Our task requires deep contextual understanding and domain-specific knowledge of logistics, and has not been explored

# Task: Extract structured logistics information from a narrative report

## Text Narrative Report

## Structured Data on the Status of Logistics Infrastructure (Statements)

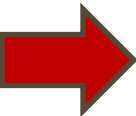
### Updates in Türkiye

#### Impact and humanitarian needs

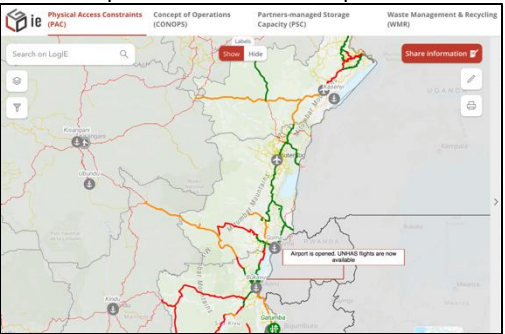
- At least 9,057, deaths and 52,979 injuries have been confirmed by the Government of Türkiye. The top three most affected districts by number of deaths are Hatay, Kahramanmaraş and Gaziantep. In Hatay alone, the number of death is as high as 3,356.
- Deaths and injuries have so far been reported in Kahramanmaraş, Gaziantep, Şanlıurfa, Diyarbakır, Adana, Adiyaman, Osmaniye, Hatay, Kilis, Malatya and Elazığ provinces.
- At least 6,444 buildings have reportedly collapsed in the country.
- As of 7 February, airports in Kahramanmaraş and Hatay remain closed due to damage. Airports in Gaziantep and Şanlıurfa are open to humanitarian flights. Airports in Malatya, Adana, Diyarbakır, Adiyaman Airports are open to flights.
- Gas flow through pipelines has been stopped in Kahramanmaraş and Gaziantep to mitigate risks of explosions.
- Schools in the affected provinces remain closed for at least one week.
- A number of key transportation routes have been impaired.
- The Government of Türkiye issued a Level 4 alarm on 6 February calling for international assistance.

#### Humanitarian response

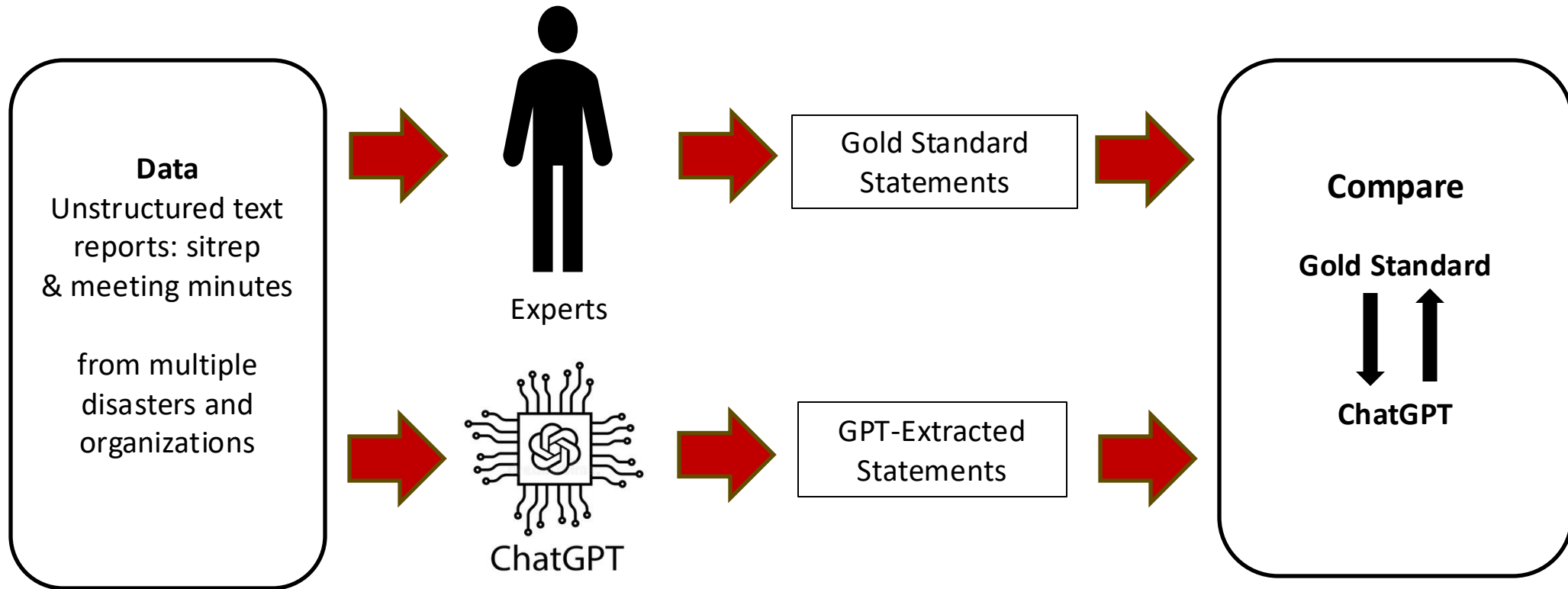
- According to AFAD, the number of search and rescue personnel in the region is 98,153 personnel, including 5,309 international personnel from 18 countries.
- UNDAC, International Search and Rescue Advisory Group (INSARAG) response teams and Emergency Medical Teams (EMT) are being mobilized to Türkiye. An UNDAC team dedicated to the response in Gaziantep arrived in Adana on 8 February with further deployments to Kahramanmaraş and potentially Adiyaman.
- More than 8,000 people have been rescued from the rubble of the buildings. Besides rescue teams, blankets, tents, food and psychological support teams were also sent to affected regions.



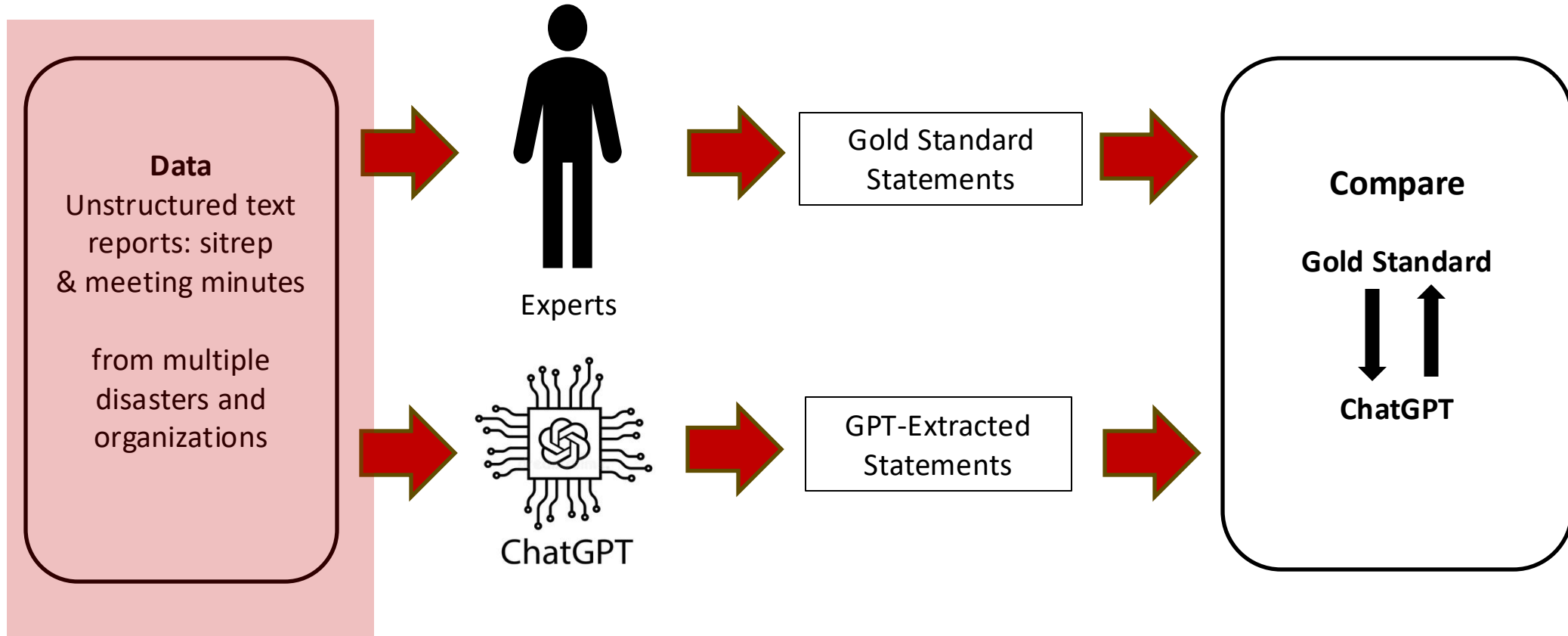
Location	Infrastructure	Status
Kahramanmaraş, Türkiye	Airport	Closed
Hatay, Türkiye	Airport	Closed
Gaziantep, Türkiye	Airport	Open
Şanlıurfa, Türkiye	Airport	Open
Malatya, Türkiye	Airport	Open
Adana, Türkiye	Airport	Open
Diyarbakır, Türkiye	Airport	Open
Adiyaman, Türkiye	Airport	Open



# Methodology Overview



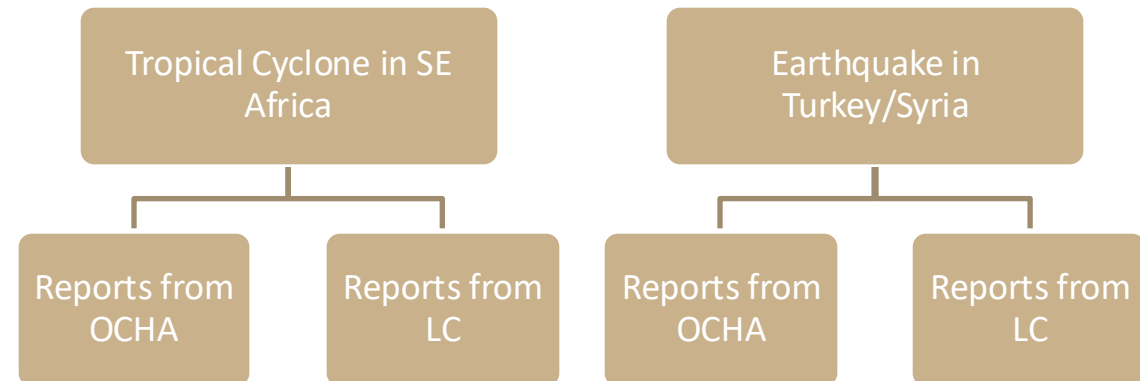
# Methodology Overview





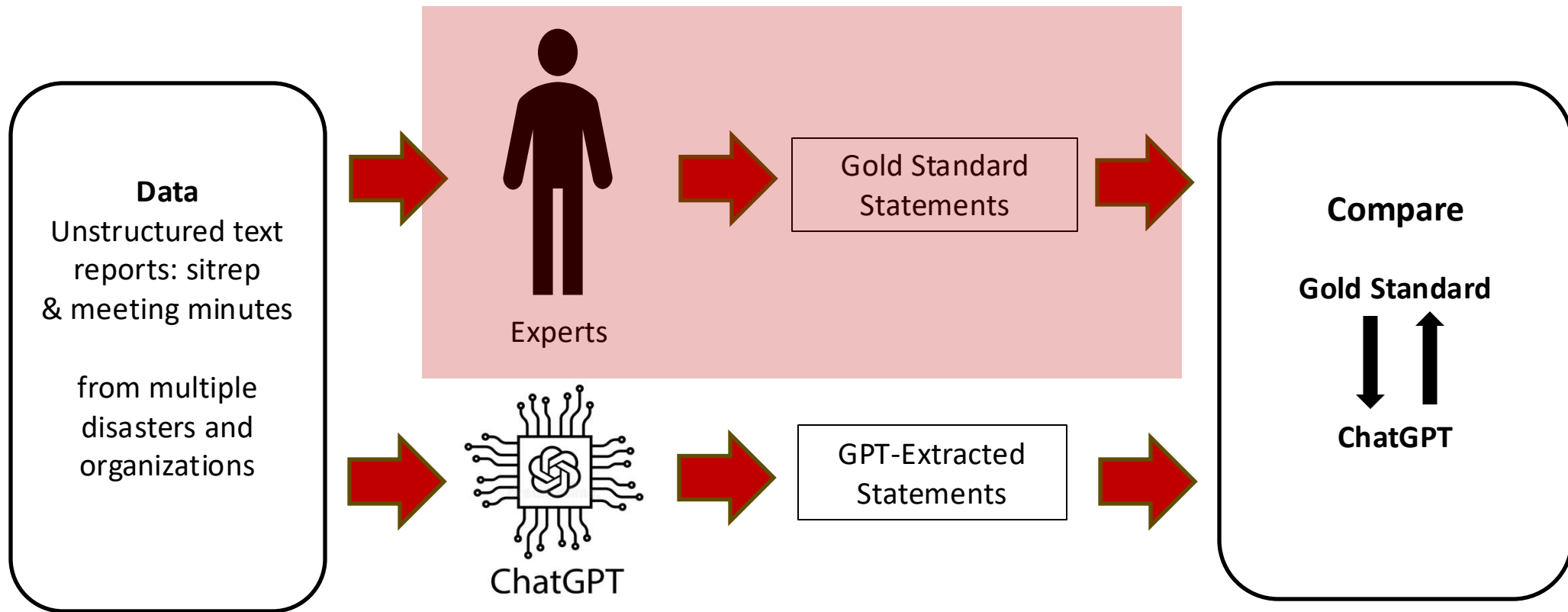
# Data: 40 reports from two disasters

- Selected **two different disasters**
  - Tropical Cyclone Freddy in Southeast Africa (SEA)
  - Earthquake in Turkey/Syria (TRKY)
- Selected **two different organizations**:
  - Broad scope: United Nations Office for the Coordination of Humanitarian Affairs (UN OCHA)
  - Narrow focus on logistics: Logistics Cluster (LC)
- Selected the **first 10 documents** produced by each organization in each disaster





# Methodology Overview



# Experts identified all key logistics information in each report

- Guidelines defined what counts as “key logistics information”
- Consistent process
  - Two researchers independently labeled multiple documents; disagreements reconciled.
- Total number of statements extracted: 396

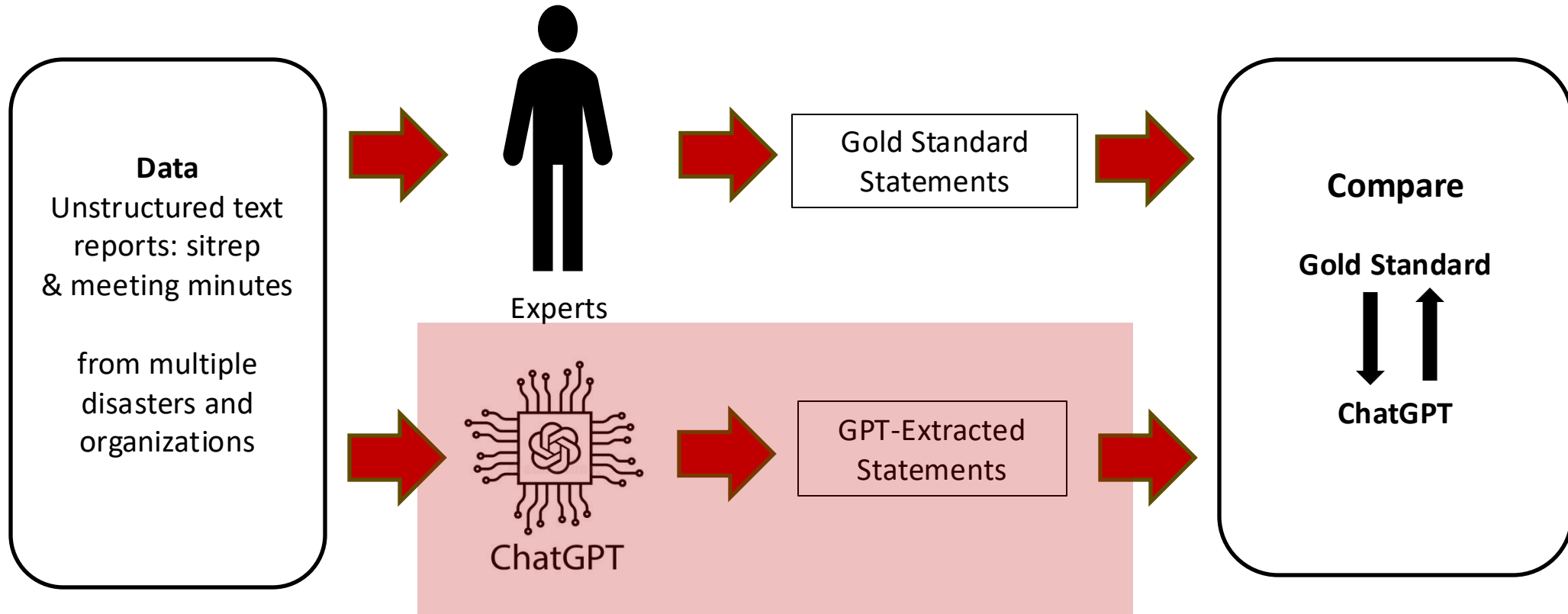
## Information to Extract

Status and location of:

- Airports
- Seaports
- Road
- Railways
- Transportation
- Storage
- Fuel
- Coordination services

Location	Infrastructure	Status
NWS	Transportation	Shortage
NWS	Fuel	Not Available
Damascus to Aleppo	Transportation	Not Available
Aleppo	Storage	Available
Homs	Storage	Available

# Methodology Overview



# ChatGPT extracts statements based on same guidelines

- ChatGPT is prompted to extract information from each document
- Prompts were iteratively refined
- Evaluated multiple model versions and temperature settings to assess variability

## Updates in Türkiye

### Impact and humanitarian needs

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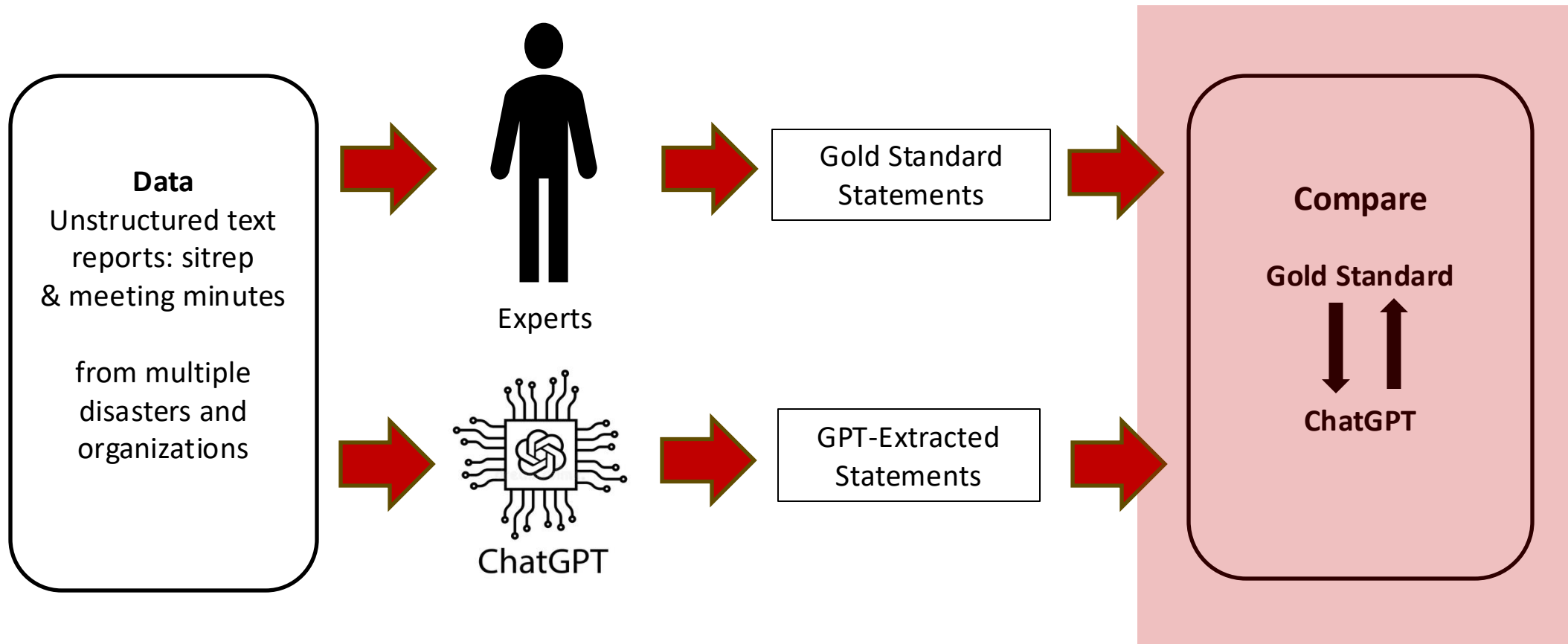
### Humanitarian response

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# Methodology Overview



# Comparison

Gold Standard			ChatGPT Extraction		
Location	Infrastructure	Status	Location	Infrastructure	Status
Bab al-Hawa	Border Crossing	Open	Bab alhawa	Entry-point	Operational



Matched  
statement

## Errors

Mis-match in statement component(s)

GPT did not catch the statement

GPT suggested an incorrect statement

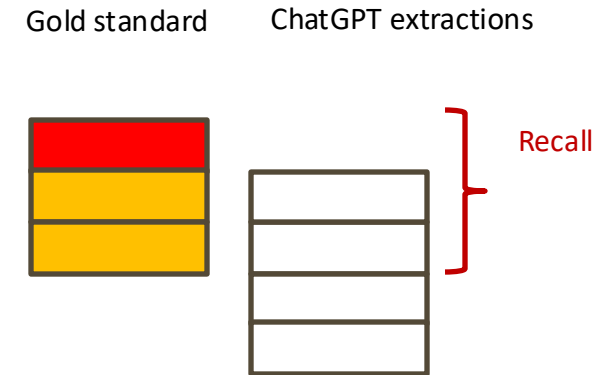
Gold Standard			ChatGPT Extraction		
Location	Infrastructure	Status	Location	Infrastructure	Status
Bab al-Hawa	Border Crossing	Open	Bab alhawa	Entry-point	Closed
Bab Salama	Border Crossing	Open			
			Border	Procedures	Controlled



# Performance Metrics

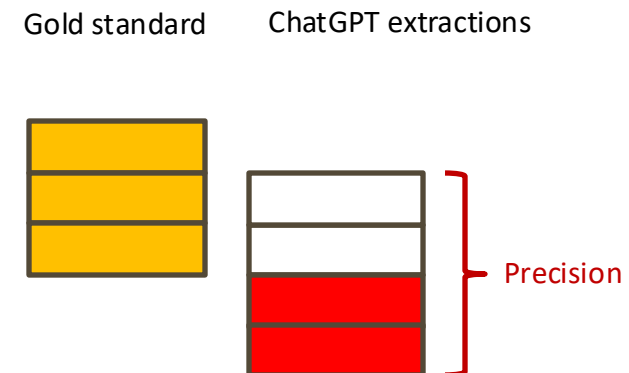
- Recall: How many of the gold standard statements does ChatGPT correctly identify?

$$\text{Recall} = \frac{\text{Matched Statements}}{\text{Gold Standard Statements}}$$



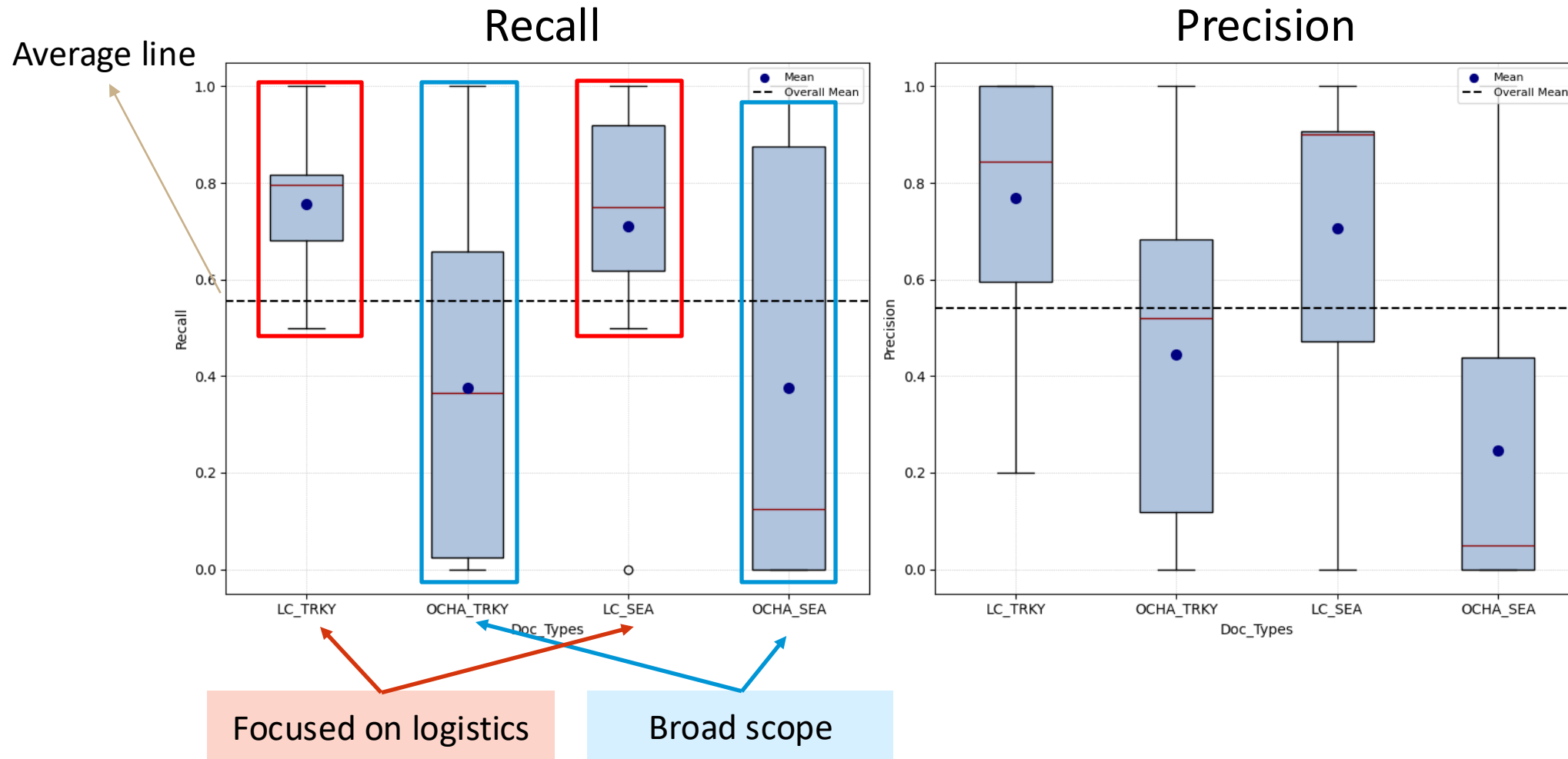
- Precision: How many of ChatGPT's statements are in the gold standard (i.e., not “extra”)?

$$\text{Precision} = \frac{\text{Matched Statements}}{\text{ChatGPT Statements}}$$



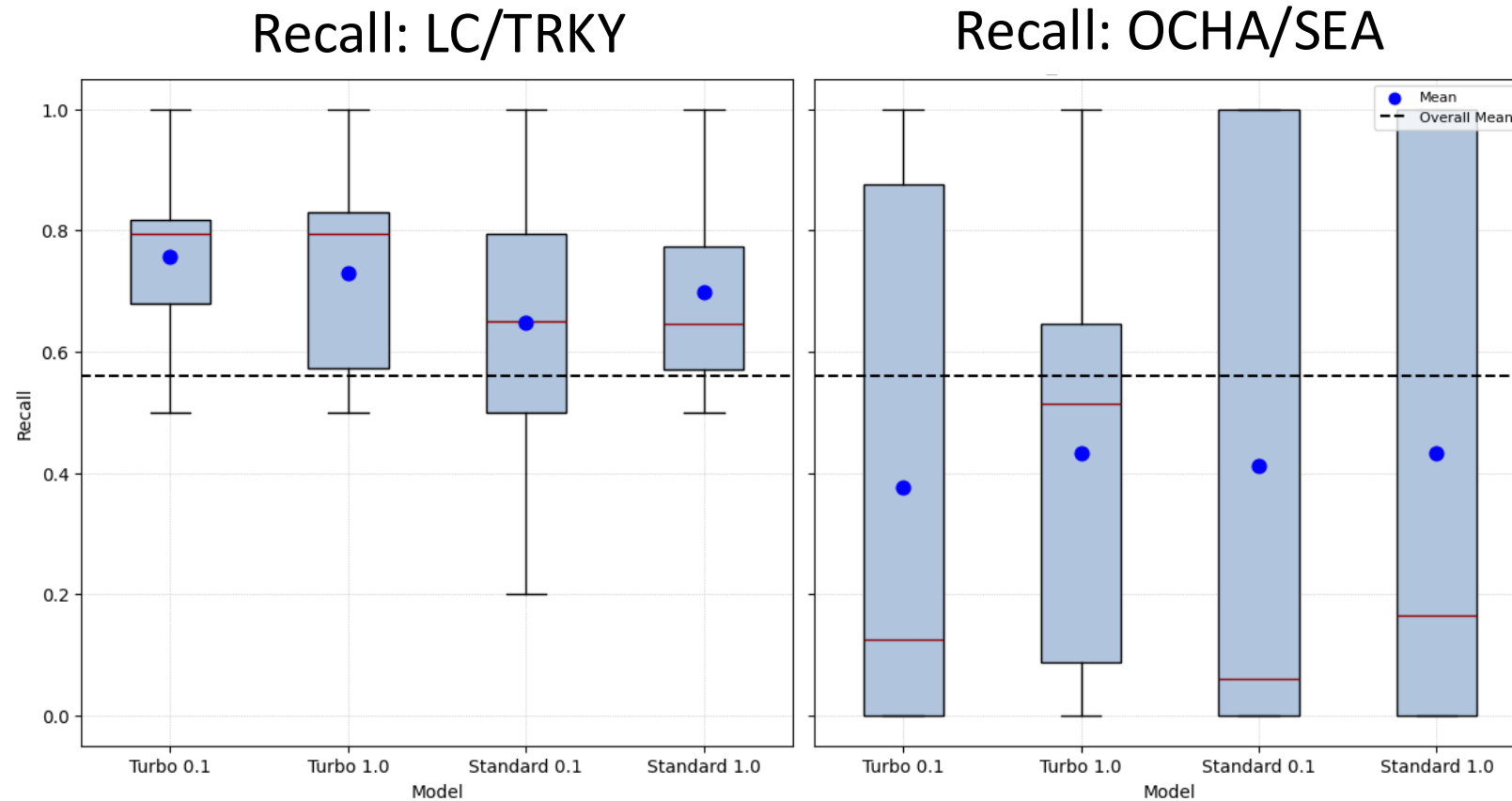
# Overall Performance

# Performance differs based on source document



Performance depends on the **organization** that produced the report

# Performance does not depend on model parameters

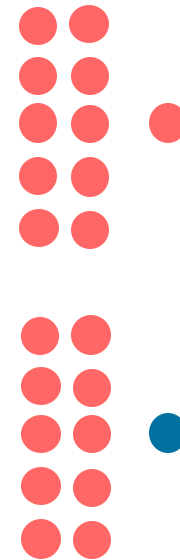


Performance is consistent across different **model versions** and **temperatures**

# Improving Performance

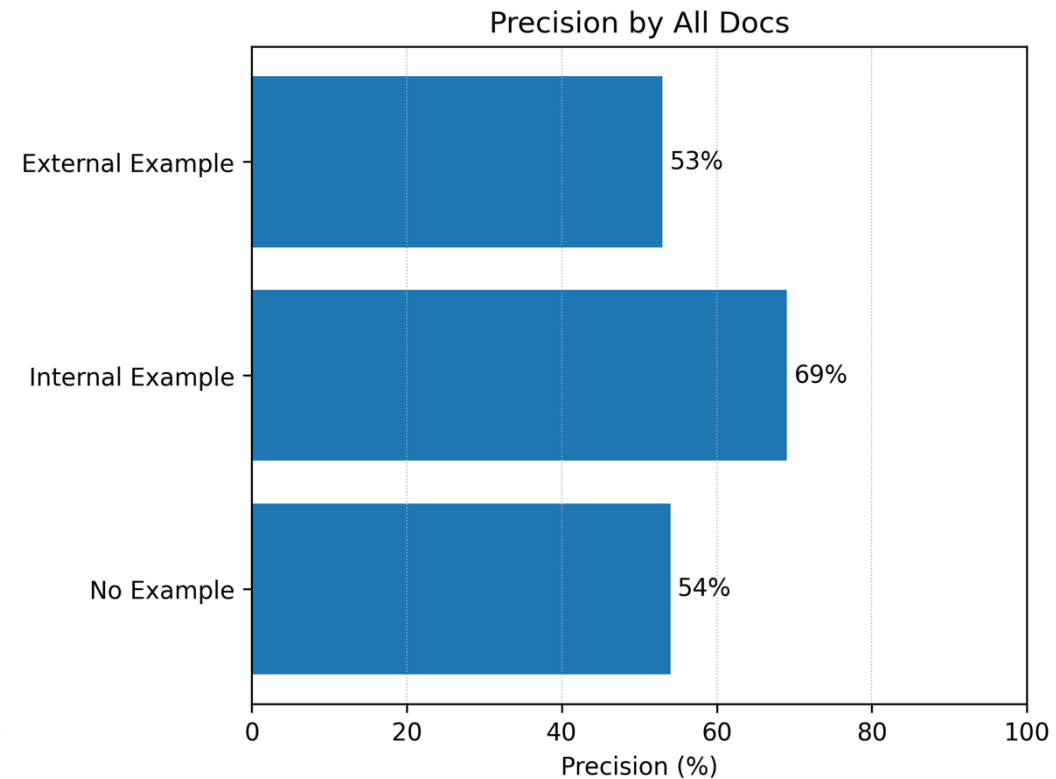
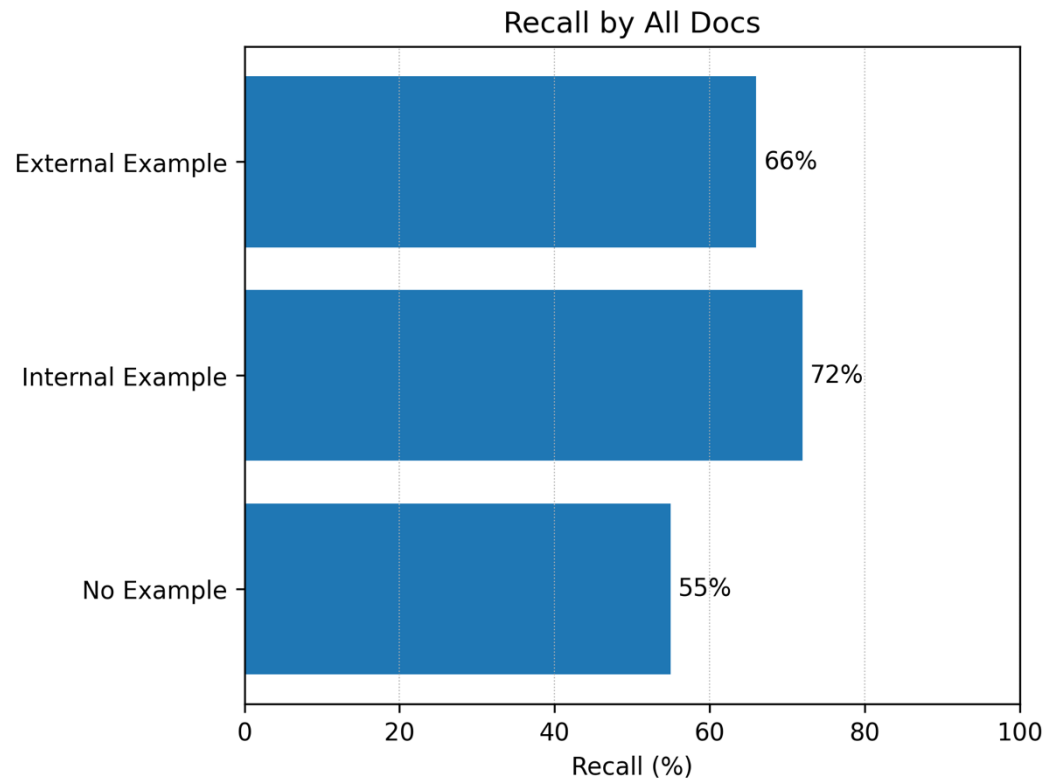
# Attempt 1: Provide an example

- Will performance improve if the prompt includes an example?
- Tested two types of examples
  - **Internal:** from a document produced by the same organization in the same disaster
  - **External:** from a document produced by a different organization in a different disaster





# Performance improves with internal examples



Presenting an example from the same dataset improves recall and precision

\*External Example: an example from different disaster and organization

\*\* Internal Example: an example from the same dataset

# Attempt 2: Remove ambiguity

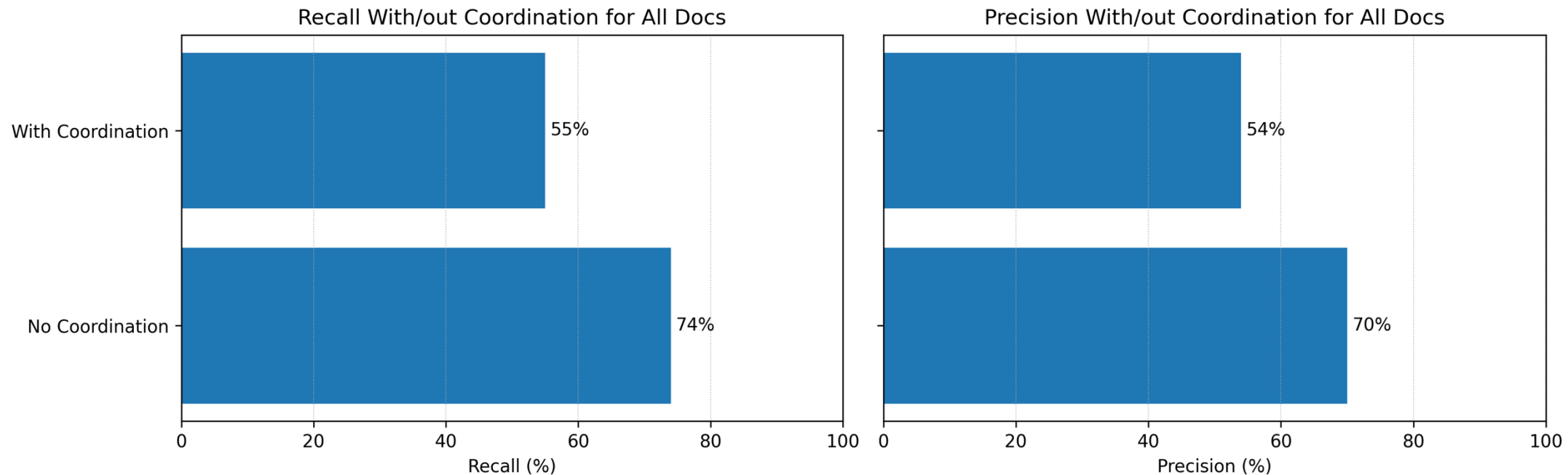
- Will performance improve if we remove ambiguous categories of information?
  - Information on coordination was very difficult to interpret in the documents
- Tested performance with coordination removed from the scope

## Information to Extract

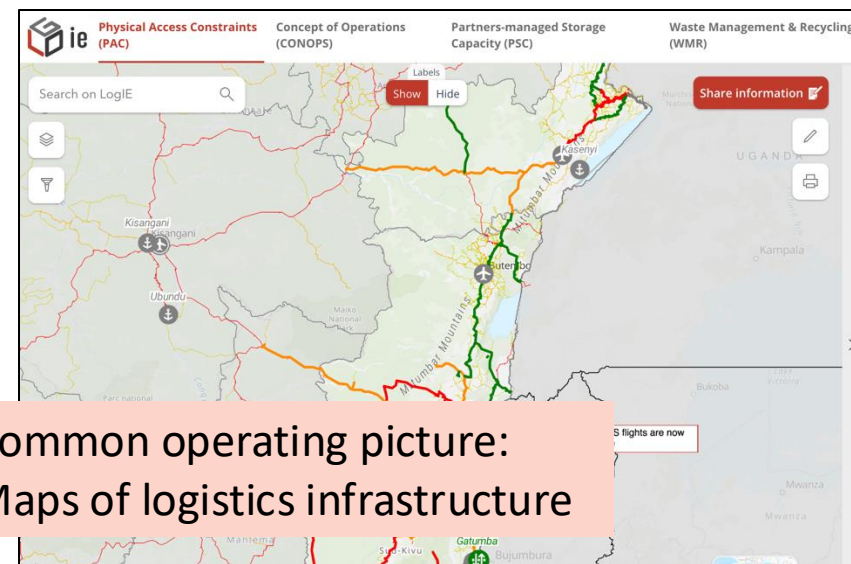
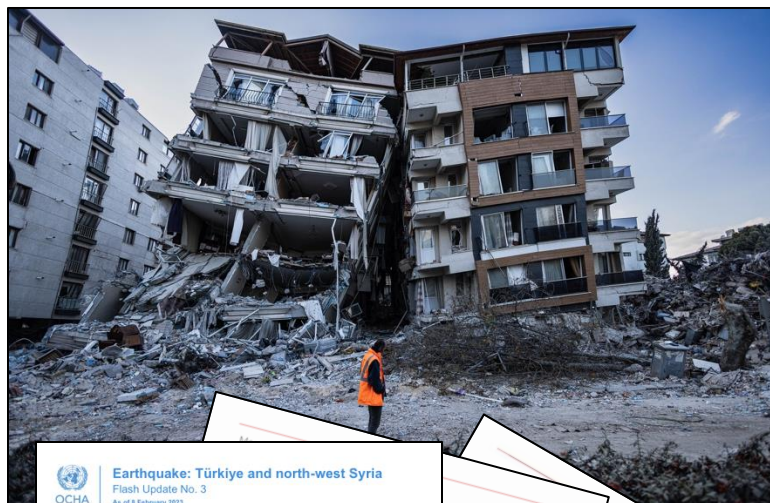
### Status and location of:

- Airports
- Seaports
- Road
- Railways
- Transportation
- Storage
- Fuel
- ~~Coordination services~~

# Performance improves when ambiguous categories are removed from consideration



# Can LLMs extract a logistics COP from narratives?



## Common operating picture: Maps of logistics infrastructure

## Unstructured narrative reports from multiple sources

# Conclusions

- Performance depends on the **documents!**
  - Works well for more direct, structured narratives (0.75-0.8 precision, recall)
  - Works less well for broader, less focused narratives (highly variable performance)
- Providing **relevant examples** improves performance
- Performance also depends on the **ambiguity** of the information to be extracted
  - Performance was poor in extracting information on coordination services, better for airports, seaports, etc.
- **Temperature and model version** has little impact on performance

# Ongoing Work

- Analyzing the nature of the errors ChatGPT makes
  - ...and how they can be fixed
- Verifying the results by feeding the model with additional information from different sources



# Thank you!

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