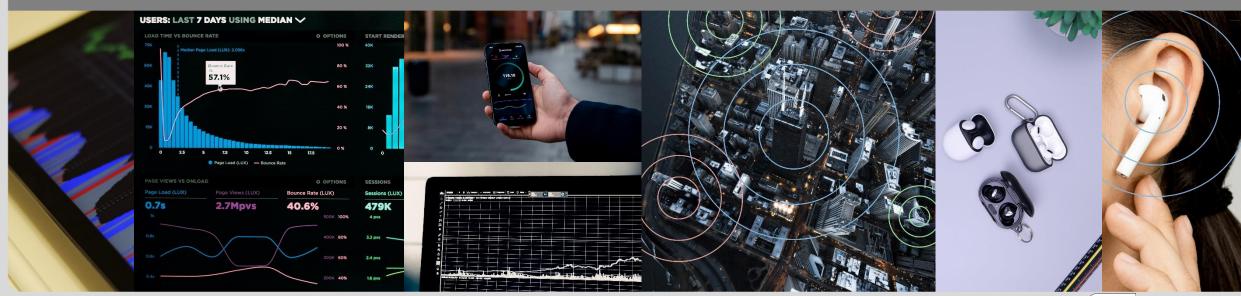




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Outlines



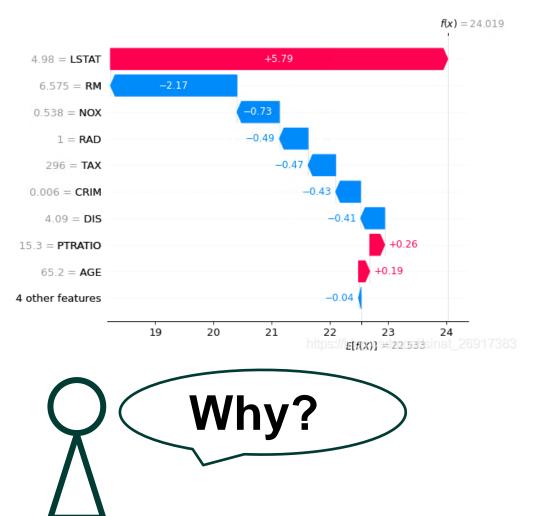
- Motivation
- Ground knowledge
- Related Works
- Example of my solution
- Realization of my solution
- Planned Evaluation



Motivation



- Problem: Current method to explain Al is not user-friendly for non-researchers.
- ■Idea: Use **ChatGPT** to explain why the feature can contribute to the classification
- ■Benefit:
 - User-friendly and easy-accessible.
 - Link features to knowledge in other fields.
- Actions: Design a **prompt** to guide GPT give the answer.





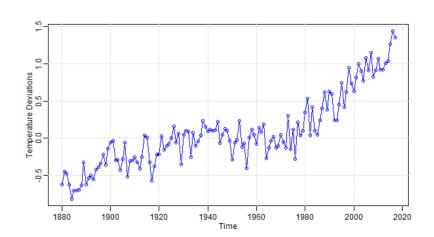
Ground knowledge



ChatGPT



Time series



Explanability



Related Works

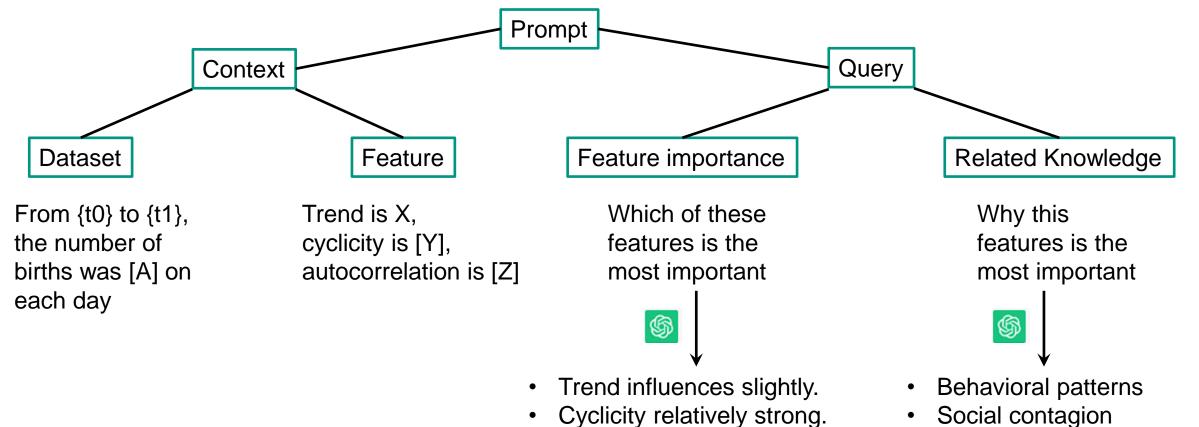


- Feature selection(tsfresh)
 - Gašper et al. provided a new method to understand the importance of time series feature generated by tsfresh
- Forecasting via Prompt
 - Xue et al. used prompt-based method to forecast time-series data
 - Set a benchmark to evaluate different methods and LLM
- Causal Inference
 - ■Stephanie et al. Proved that GPT can generate causal graphs in a high accuracy.

Example of my solution

Daily total female births in California





Autocorrelation significantly

Reporting and registration

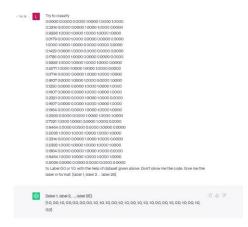
Medical scheduling

processes

Realization of my solution



- Feature selection(tsfresh)
- Prompt design
 - Context
 - Dataset description
 - Features
 - Query
 - Study which key words improve answer stability









- Combine feature importance to domain knowledge
 - Get key words from answer and ask automatically.
- Prove the reliability



Planned Evaluation



- Research object: A time-series dataset (From UCR)
 - With explanation generated by other methods
- ■Target:
 - Stable desired answer
 - ■Better prompt, avoiding incorrect answer or unwished format
 - Prove to be reliable
 - Compare GPT-selected importance and pre-set importance
 - Modify original data to strongly affect feature
 - ■Test consistency

Prompt engineering



- 1. Write clear and specific instructions
 - 1.使用分隔符
 - 2.获得指定输出 如JSON
 - 3.让LLM检查文本是否满足条件
 - 4.为模型提供几个问答的例子,让 LLM 熟悉这种风格
- 2. Give the model time to think
 - 1.明确完成任务所需的步骤: Step1, Step2
 - 2.获取结构化的输出:
- 3.LLM的具体功能:
 - 1.总结,推理(态度,情感,主题),转换,扩展
- 4.迭代优化:
 - 1.更明确地聚焦于某个方面:

Bsp:The description is intended for furniture retailers, so should be technical in nature and focus on the materials the product is constructed from.

- Use the following format:
- Text: <text to summarize>
- Summary: <summary>
- Translation: <summary translation>
- Names: <list of names in French summary>
- Output JSON: <json with summary and num_names>

Dataset

Acuteinflammation & acutenephritis



- Method: 15 times, 25 tests, decision tree vs ChatGPT
 - ■Decision tree: 100% correct on both datasets
 - ChatGPT: big range, not stable
 - ■Old_Prompt: There are A acuteinflammation, and each has B columns. Each column C is ...

 Try to classify following... to Label 0.0 or 1.0, with the help of dataset given above.

Don't show me the code. Give me the label in format: [label 1, label 2, ..., label D].

New_Prompt: You are a data analyst, your job is to classify the given time series dataset.

There are 93 acutenephritis, and each has 6 columns.

The dataset will be given in format: [column 1, column 2, ..., column 6, label].

False rate:

Acuteinflammation: 20.0%, 16.0%, 20.0%, 20.0%, 20.0%, 36.0%, 20.0%, 28.0%, 16.0%, 20.0%, 16.0%, 16.0%, 20.0%, 28.0%, 20.0%

Acutenephritis: 40.0%, 40.0%, 12.0%, 40.0%, 8.0%, 36.0%, 36.0%, 28.0%, 44.0%, 40.0%, 32.0%, 36.0%, 8.0%, 24.0%, 0.0%

Average: 21.06%(35.2%), 28.26%(40.8%)

Dataset

Acuteinflammation & acutenephritis



Repr2Seq



- ■使用14.01.2019-12.01.2022的股票TS
- ■使用了Encoder-decoder模型,来描述TS结构和关键信息
 - ■Encoder分为Sequence Processor+TS2Vec
 - ■SP: $q_i = \{x_{i_1} x_{i_1}, ..., x_{i_j} x_{i_1}, ..., x_{i_m} x_{i_m}\}$
 - ■TS2Vec: 计算Loss
 - Decoder:
 - ■用argmax operation收集了生成词的目录
- ■Repr2Seq相比Transformer和Seq2seq有着更高的得分(BLEU)

Generative Agent



Memory and retrieval

- takes the agent's current situation as input and returns a subset of the memory stream
- \blacksquare score = α recency · recency + α importance · importance + α relevance · relevance.

Reflection

- when the sum of the importance scores for the latest events exceeds a certain threshold
- Agents generate trees of reflections
- Planning and Reacting
 - Reacting and Updating Plans + Dialogue

Causal inference



- Can Large Language Models Build Causal Graphs
 - Using medical context
 - Determining whether GPT-3 can signal the presence or absence of an edge between two variables in a directed acyclic graph from the medical context.
 - Evaluating whether the use of certain language in prompts or linking verbs improves the classification accuracy of GPT-3.
 - Normally, a better prompt after iteration performs better
 - ■Both verb and noun
 - Specificity lowers accuracy

