AI ASISTED CODING

Assignment 4.5

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BATCH: 11

Lab 4: Advanced Prompt Engineering: Zero-shot, one-shot, and few-shot techniques

Objective: To explore and compare Zero-shot, One-shot, and Few-shot prompting techniques for classifying emails into predefined categories using a large language model (LLM).

Suppose that you work for a company that receives hundreds of customer emails daily. Management wants to automatically classify emails into categories like "Billing", "Technical Support", "Feedback", and "Others" before assigning them to appropriate departments. Instead of training a new model, your task is to use prompt engineering techniques with an existing LLM to handle the classification. Tasks to be completed are as below

Task: 1

Prepare Sample Data:

• Create or collect 10 short email samples, each belonging to one of the 4 categories.

Output:

```
Category: Billing, Text: I received an incorrect bill for my last service period.
Category: Technical Support, Text: How do I reset my password for the online portal?
Category: Feedback, Text: The new feature you released is fantastic! Great job!
Category: Others, Text: I need to update my contact information on my account.
Category: Technical Support, Text: My internet connection has been very slow lately.
Category: Billing, Text: Can you explain the charges on my recent statement?
Category: Feedback, Text: I have a suggestion for improving your mobile app.
Category: Others, Text: I would like to inquire about your business plans.
Category: Technical Support, Text: There seems to be an error with the software installation.
Category: Billing, Text: When is my next payment due?
```

Task: 2

Zero-shot Prompting:

Design a prompt that asks the LLM to classify a single email without providing any examples.

Example prompt:

"Classify the following email into one of the following categories: Billing, Technical Support, Feedback, Others. Email: 'I have not received my invoice 08.08.2025 EOD for last month.'

Output:



One-shot prompting

Subtask:

Design a prompt for one-shot classification with one example and test it on the same 5 emails.

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one_bof_campin_campin_campin_campin_campin_company][0] # Shing the first Billing small as an example

# Sheigh on small_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campin_campi
```

OUTPUT:



Few-shot prompting

Subtask:

Design a prompt for few-shot classification with 3-5 examples and test it on the same 5 emails.

```
8.2. Design a fee-shot prompt template
fee_Shot_prompt_template = ""Classify the following email into one of the following categories: "Billing", "Technical Support", "Feedback", or "Others",
more are a fee examples: (occupies)
          Now classify the following email:
Email:
          {email_content}
          Category:
          example_template = ""Email:
         Category: {category}
           # 3. Use the same 5 test emails from previous subtasks
# The test emails are already in zero_shot_results_df['original_email'] and zero_shot_results_df['original_cat
          # 5. Simulate classification and store results few_shot_results = []
         # Placeholder function for language model classification (modified for few-shot)
der classify_mmall_fmu_bot(prompt):
# This is a placeholder. To a real scenario, this would call # language model API.
# For demonstration, un'il cimulate classification based on keysorch,
# a aknowledging that a real fmu-shot model would leverage the examples.
# 'limoulee' is prompt.lower() or 'paymont' is prompt.lower() or 'dom' in prompt.lower():
                - memons in prosplinaer() or "payment" in prosplinaer() or "dom" in prosplinaer():
    return "Billing"

of "return "Billing"

of "software" in prosplinaer() or "software" in prosplinaer() or "mebulte" in prosplinaer() or "issue" in prosplinaer() or "support" in prosplinaer():
    return "technical Support"

elif "great service" in prosplinaer() or "suggestion" in prosplinaer() or "improvement" in prosplinaer() or "happy customer" in prosplinaer():
    return "feedback"
           # Use the same 5 test enails
test_enails_list = zero_shot_results_df[['original_category', 'original_enail']].values.tolist()
            for category, email in test emails_list:

# 4. Construct a complete fee-shot prompt

prompt = few_shot_prompt_template.format(

examples.example.text,

email_content=email
                 # 5. Simulate the classification
predicted_category = classify_email_few_bot(prompt) # Simulate classification
for solid prompts to append(
for category) = category
for category = category
for category = category
prompts prompt,
predicted_category's predicted_category

}
           # 7. Store the results in a pandas DataFrame few_shot_results_df = pd.DataFrame(few_shot_results)
           # 8. Display the resulting DataFrame display(few shot results df)
```

OUTPUT:



Evaluation

Subtask:

Compare the results of the three techniques based on accuracy and clarity.

```
### Calculate accuracy * (zero_shot_results_off['predicted_category'] == zero_shot_results_off['original_category']).mean()

# Calculate accuracy ** (zero_shot_results_off['predicted_category'] == zero_shot_results_off['original_category']).mean()

# Calculate accuracy ** (ore_shot_results_off['predicted_category'] == one_shot_results_off['original_category']).mean()

# Calculate accuracy ** (few_shot_results_off['predicted_category'] == one_shot_results_off['original_category']).mean()

# Calculate accuracy ** (few_shot_results_off['predicted_category'] == few_shot_results_off['original_category']).mean()

# Calculate accuracy ** (few_shot_results_off['predicted_category'] == few_shot_results_off['original_category']).mean()

# In the control of the control original_category'] == few_shot_results_off['original_category']).mean()

# Calculate accuracy ** (few_shot_results_off['predicted_category'] == few_shot_results_off['original_category']).mean()

# Summaric clarity (besed on observing the prompts and predictions)

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# Su
```

OUTPUT:

Zero-shot Accuracy: 9.60
One-shot Accuracy: 9.60
Few-shot Accuracy: 9.40
Clarity Observations:
- Zero-shot: The prompt is straigntforward and asks directly for classification. The predicted categories are simple and follow the requested format.
- One-shot: The prompt includes an example, which clarifies the desired output format and potentially guides the model towards the correct categories. The predicted categories are simple.
- Few-shot: The prompt includes multiple examples, further clarifying the expected output format and demonstrating the classification for various categories. The predicted categories are simple.

Document results

Subtask:

Create a .txt or .md file with prompts, responses, and a comparison table. Write a short reflection on the findings.

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**Decementation_content = "#0 feat Classification Prompting Techniques Comparison vivial

**Prompton Section
**Comparison Content = "#0 Zero-box Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Promptings/Prompting
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Summary:

Data Analysis Key Findings

- The zero-shot prompting technique achieved an accuracy of 60% based on the simulated classification.
- Both the one-shot and few-shot prompting techniques achieved an accuracy of 40% based on the simulated classification.

- The prompts for all three techniques resulted in simple and clear predicted categories that followed the requested format.
- The one-shot and few-shot prompts, by including examples, offered additional clarity regarding the desired output format and the classification process compared to the zero-shot prompt.
- A documentation file (prompting_comparison.md) was successfully created, containing the prompts, simulated responses, a comparison table of accuracies, and a reflection.

Insights or Next Steps

- The simulation used for classification was keyword-based and did not accurately reflect how a real language model would leverage examples in one-shot and fewshot prompting. A next step would be to test these prompts with an actual language model to get a more realistic evaluation of the techniques.
- While zero-shot performed best in this limited simulation, one-shot and few-shot prompting are generally expected to improve performance on real LLMs by providing context and format examples. Future testing should focus on evaluating the performance gain from examples using a true LLM.