

## **LAB ASSIGNMENT – 4.4**

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### **1. Sentiment Classification for Customer Reviews**

#### **Scenario:**

An e-commerce platform wants to analyze customer reviews and classify them into Positive, Negative, or Neutral sentiments using prompt engineering.

#### **Tasks:**

- a) Prepare 6 short customer reviews mapped to sentiment labels.
- b) Design a Zero-shot prompt to classify sentiment.
- c) Design a One-shot prompt with one labeled example.
- d) Design a Few-shot prompt with 3–5 labeled examples.
- e) Compare the outputs and discuss accuracy differences.

#### **Scenario**

An e-commerce platform wants to classify customer reviews as **Positive, Negative, or Neutral** using prompt engineering.

#### **(a) Sample Customer Reviews**

| <b>Review</b>   | <b>Sentiment</b> |
|---|------------------|
| “The product quality is excellent and delivery was fast.” | Positive         |
| “Very satisfied with the purchase.”                       | Positive         |
| “The item is okay but nothing special.”                   | Neutral          |
| “Delivery was delayed but product is fine.”               | Neutral          |
| “Worst experience ever, totally disappointed.”            | Negative         |
| “The product stopped working in two days.”                | Negative         |

#### **AI Generated Code:**

#### **(b) Zero-Shot Prompt**

Classify the sentiment of the following review as Positive, Negative, or Neutral:

"The product quality is excellent and delivery was fast."

```
Classify the sentiment of the following review as Positive, Negative, or Neutral:  
  
"The product quality is excellent and delivery was fast."
```

**Output:** Positive

#### **(c) One-Shot Prompt**

Example:

Review: "The product is terrible and useless."

Sentiment: Negative

Now classify:

Review: "The product quality is excellent and delivery was fast."

**Example:**

Review: "The product is terrible and useless."

Sentiment: Negative

Now classify:

Review: "The product quality is excellent and delivery was fast."

**Output:** Positive

#### (d) Few-Shot Prompt

Review: "The product is amazing and works perfectly."

Sentiment: Positive

Review: "Delivery was late and the box was damaged."

Sentiment: Negative

Review: "It's okay, not too good or bad."

Sentiment: Neutral

Now classify:

Review: "Very satisfied with the purchase."

Review: "The product is amazing and works perfectly."

Sentiment: Positive

Review: "Delivery was late and the box was damaged."

Sentiment: Negative

Review: "It's okay, not too good or bad."

Sentiment: Neutral

Now classify:

Review: "Very satisfied with the purchase."

**Output:** Positive

#### (e) Comparison & Analysis

##### Method Accuracy Reason

Zero-shot Medium No examples provided

One-shot Better Model understands pattern

Few-shot Best Context improves classification

Few-shot prompting gives the most accurate and consistent results.

## 2. Email Priority Classification

### Scenario:

A company wants to automatically prioritize incoming emails into High Priority, Medium Priority, or Low Priority.

### Tasks:

1. Create 6 sample email messages with priority labels.
2. Perform intent classification using Zero-shot prompting.
3. Perform classification using One-shot prompting.
4. Perform classification using Few-shot prompting.
5. Evaluate which technique produces the most reliable results and why.

### Sample Emails

| Email                                   | Priority |
|---|----------|
| "Server is down, fix immediately!"      | High     |
| "Client meeting postponed to tomorrow." | Medium   |
| "Thank you for your support."           | Low      |
| "Urgent payment issue, respond ASAP."   | High     |
| "Weekly report attached."               | Medium   |
| "Just checking in."                     | Low      |

### Zero-Shot Prompt

Classify this email as High, Medium, or Low priority:

"Server is down, fix immediately!"

```
Classify this email as High, Medium, or Low priority:  
"Server is down, fix immediately!"
```

Output: High

### One-Shot Prompt

Email: "Payment failed, urgent fix needed."

Priority: High

Now classify:

"Client meeting postponed to tomorrow."

```
Email: "Payment failed, urgent fix needed."  
Priority: High  
  
Now classify:  
"Client meeting postponed to tomorrow."
```

Output: Medium

### Few-Shot Prompt

Email: "System crash reported."

Priority: High

Email: "Weekly update attached."

Priority: Medium

Email: "Thanks for your help."

Priority: Low

Now classify:

"Urgent payment issue, respond ASAP."

Email: "System crash reported."

Priority: High

Email: "Weekly update attached."

Priority: Medium

Email: "Thanks for your help."

Priority: Low

Now classify:

"Urgent payment issue, respond ASAP."

Output: High

## Evaluation

### Method Reliability

Zero-shot Moderate

One-shot Good

Few-shot Best

Few-shot gives context → better accuracy

## 3. Student Query Routing System

### Scenario:

A university chatbot must route student queries to Admissions, Exams, Academics, or Placements.

### Tasks:

1. Create 6 sample student queries mapped to departments.
2. Implement Zero-shot intent classification using an LLM.
3. Improve results using One-shot prompting.
4. Further refine results using Few-shot prompting.
5. Analyze how contextual examples affect classification accuracy.

### Departments

- Admissions
- Exams
- Academics
- Placements

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## Sample Queries

| Query                            | Department |
|----------------------------------|------------|
| "What is the admission process?" | Admissions |
| "When are semester exams?"       | Exams      |
| "Explain Python syllabus."       | Academics  |
| "Placement companies list?"      | Placements |
| "Fee structure details"          | Admissions |
| "Result declaration date?"       | Exams      |

### Zero-Shot Prompt

Classify the query:

"When are semester exams?"

**Classify the query:**

**"When are semester exams?"**

Output: Exams

### One-Shot Prompt

Query: "How to apply for admission?"

Department: Admissions

Now classify:

"When are semester exams?"

**Query: "How to apply for admission?"**

**Department: Admissions**

**Now classify:**

**"When are semester exams?"**

Output: Exams

### Few-Shot Prompt

Query: "Syllabus for AI subject?"

Department: Academics

Query: "Placement drive details?"

Department: Placements

Query: "Fee structure?"

Department: Admissions

Now classify:

"When will results be announced?"

```
Query: "Syllabus for AI subject?"  
Department: Academics  
  
Query: "Placement drive details?"  
Department: Placements  
  
Query: "Fee structure?"  
Department: Admissions  
  
Now classify:  
"When will results be announced?"
```

Output: Exams

### Analysis

Few-shot prompting gives better domain understanding and routing accuracy.

## 4. Chatbot Question Type Detection

### Scenario:

A chatbot must identify whether a user query is Informational, Transactional, Complaint, or Feedback.

### Tasks:

1. Prepare 6 chatbot queries mapped to question types.
2. Design prompts for Zero-shot, One-shot, and Few-shot learning.
3. Test all prompts on the same unseen queries.
4. Compare response correctness and ambiguity handling.
5. Document observations.

### Question Types

- Informational
- Transactional
- Complaint
- Feedback

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### Sample Queries

| Query                         | Type          |
|-------------------------------|---------------|
| “What is AI?”                 | Informational |
| “Book a ticket for tomorrow.” | Transactional |
| “The app crashes often.”      | Complaint     |
| “Great service, thank you!”   | Feedback      |
| “How to reset password?”      | Informational |
| “Refund my payment.”          | Transactional |

### Prompting Comparison

**Zero-Shot:**

Classify the query: "The app crashes often."

```
Classify the query: "The app crashes often."
```

→ Complaint

**One-Shot:**

Query: "Great service!"

Type: Feedback

Now classify:

"The app crashes often."

```
Query: "Great service!"
```

```
Type: Feedback
```

```
Now classify:
```

```
"The app crashes often."
```

→ Complaint

**Few-Shot:**

Query: "What is AI?"

Type: Informational

Query: "Book my movie tickets."

Type: Transactional

Query: "The app is not working properly."

Type: Complaint

Query: "Excellent customer support!"

Type: Feedback

Now classify:

Query: "I want to cancel my order."

```
Query: "What is AI?"  
Type: Informational  
  
Query: "Book my movie tickets."  
Type: Transactional  
  
Query: "The app is not working properly."  
Type: Complaint  
  
Query: "Excellent customer support!"  
Type: Feedback  
  
Now classify:  
Query: "I want to cancel my order."
```

→ Transactional

### Comparison Table

#### Method Accuracy Reason

|           |        |                          |
|-----------|--------|--------------------------|
| Zero-shot | Medium | No learning examples     |
| One-shot  | Good   | Learns from one pattern  |
| Few-shot  | Best   | Learns multiple patterns |

### Observations

- ✓ Few-shot prompting gives **highest accuracy**
- ✓ Context improves intent understanding
- ✓ Reduces ambiguity
- ✓ Best suited for real-world chatbot systems

### Conclusion

Few-shot prompting is the **most reliable method** for chatbot question classification because:

- It understands patterns better
- Handles ambiguous queries
- Produces consistent results
- Suitable for production chatbots

## 5. Emotion Detection in Text

### Scenario:

A mental-health chatbot needs to detect emotions: Happy, Sad, Angry, Anxious, Neutral.

### Tasks:

1. Create labeled emotion samples.
2. Use Zero-shot prompting to identify emotions.

3. Use One-shot prompting with an example.
4. Use Few-shot prompting with multiple emotions.
5. Discuss ambiguity handling across techniques.

### Scenario

A mental-health chatbot must detect user emotions in order to respond appropriately.

The emotions to be identified are:

- Happy
- Sad
- Angry
- Anxious
- Neutral

### Sample Emotion Data

| Text                                 | Emotion |
|--------------------------------------|---------|
| "I am very happy today!"             | Happy   |
| "I feel lonely and depressed."       | Sad     |
| "This is unacceptable and annoying!" | Angry   |
| "I am worried about my exams."       | Anxious |
| "Okay, noted."                       | Neutral |
| "Everything is going well."          | Happy   |

### Zero-Shot Prompt

#### Prompt

Detect the emotion in the following sentence:

"I am worried about my exams."

**Detect the emotion in the following sentence:**

**"I am worried about my exams."**

#### Output

Anxious

### One-Shot Prompt

#### Prompt

Example:

Text: "I feel very sad today."

Emotion: Sad

Now classify:

Text: "I am worried about my exams."

**Example:**

Text: "I feel very sad today."

Emotion: Sad

**Now classify:**

Text: "I am worried about my exams."

**Output**

Anxious

**Few-Shot Prompt**

**Prompt**

Text: "I am very happy today."

Emotion: Happy

Text: "This is extremely frustrating."

Emotion: Angry

Text: "I feel nervous about my interview."

Emotion: Anxious

Text: "Nothing special, just normal."

Emotion: Neutral

Now classify:

Text: "I am worried about my exams."

Text: "I am very happy today."

Emotion: Happy

Text: "This is extremely frustrating."

Emotion: Angry

Text: "I feel nervous about my interview."

Emotion: Anxious

Text: "Nothing special, just normal."

Emotion: Neutral

Now classify:

Text: "I am worried about my exams."

**Output**

Anxious

## Comparison & Analysis

### Method Accuracy Remarks

|           |        |                         |
|-----------|--------|-------------------------|
| Zero-shot | Medium | No context provided     |
| One-shot  | Good   | Learns from one example |
| Few-shot  | Best   | Handles emotion clearly |

### Observation

- Few-shot prompting gives **highest accuracy**
- Emotional ambiguity is handled better
- Useful for mental-health chatbots
- Improves response quality and reliability

### Conclusion

Few-shot prompting is the **most effective method** for emotion detection because it:

- Provides context
- Improves classification accuracy
- Reduces ambiguity
- Works well in real-world chatbot applications