

# **AI ASSISTANT CODING**

## **ASSIGNMENT 4.1**

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### **LAB-OBJECTIVES**

To explore and apply different levels of prompt examples in AI-assisted code generation.

To understand how zero-shot, one-shot, and few-shot prompting affects AI output quality.

To evaluate the impact of context richness and example quantity on AI performance

### **Q1:- CUSTOMER EMAIL SATISFACTION**

A company receives a large number of customer emails every day and wants to automatically classify them into the following categories:

- Billing
- Technical Support
- Feedback
- Others

Instead of training a new machine learning model, the company decides to use prompt engineering techniques with an existing large language model.

## **Task 1**

### TASKS

1. Prepare five short sample emails, each belonging to one of the above categories.
2. Write a zero-shot prompt to classify a given email into one of the categories without providing any examples.
3. Write a one-shot prompt by including one labeled email example and ask the model to classify a new email.
4. Write a few-shot prompt by including two or three labeled email examples and ask the model to classify a new email.
5. Compare the outputs obtained using zero-shot, one-shot, and few-shot prompting techniques and briefly comment on their effectiveness.

### **PROMPT**

Doc String:

'''

use keywords to classify  
dont use nlp for classification  
1. Sample data creation  
create a sample for emails of 5  
in following categories  
Billing, Technical Support, Feedback, Other

2. Zero Shot Prompting  
classify a given email into one of the given categories

3. One Shot Prompting

Example:

" Your Credit card bill is due on the 15th of this month."

Category: Billing

Now classify the following email into Billing, Technical Support, Feedback, or Other.

4. Few Shot Prompting

Example 1:

" I haven't been able to connect to the internet since yesterday."

Category: Technical Support

Example 2:

" Received my order late, not satisfied with the service."

Category: Feedback

## 5. Compare Zero-shot, One-shot, and Few-shot Prompting

Review the performance of each prompting technique on the same set of emails.

...

### CODE:

```
1 import random
2
3 # Sample data creation
4 emails = [
5     ("Your Credit card bill is due on the 15th of this month.", "Billing"),
6     ("I am unable to connect to the internet since yesterday.", "Technical Support"),
7     ("Received my order late, not satisfied with the service.", "Feedback"),
8     ("Can you provide more details about your product?", "Other"),
9     ("I need help with resetting my account password.", "Technical Support")
10 ]
11
12 categories = ["Billing", "Technical Support", "Feedback", "Other"]
13
14 # Zero Shot Prompting
15 def zero_shot_classification(email):
16     if "bill" in email.lower() or "credit card" in email.lower():
17         return "Billing"
18     elif "connect" in email.lower() or "help" in email.lower() or "password" in email.lower():
19         return "Technical support"
20     elif "order" in email.lower() or "service" in email.lower():
21         return "Feedback"
22     else:
23         return "Other"
24
25 # One Shot Prompting
26 def one_shot_classification(email):
27     example = "Your Credit card bill is due on the 15th of this month."
28     example_category = "Billing"
29     if example_category in zero_shot_classification(email):
30         return example_category
31     return zero_shot_classification(email)
```

```

33 # Few Shot Prompting
34 def few_shot_classification(email):
35     examples = [
36         ("I am unable to connect to the internet since yesterday.", "Technical Support"),
37         ("Received my order late, not satisfied with the service.", "Feedback"),
38         ("Your Credit card bill is due on the 15th of this month.", "Billing")
39     ]
40     for example_email, category in examples:
41         if any(word in email.lower() for word in example_email.lower().split()):
42             return category
43     return zero_shot_classification(email)
44
45 # Compare Zero-shot, One-shot, and Few-shot Prompting
46 def compare_prompting_techniques(emails):
47     results = {"Zero Shot": [], "One Shot": [], "Few Shot": []}
48     for email, actual_category in emails:
49         results["Zero Shot"].append((email, zero_shot_classification(email), actual_category))
50         results["One Shot"].append((email, one_shot_classification(email), actual_category))
51         results["Few Shot"].append((email, few_shot_classification(email), actual_category))
52     return results
53
54 # Evaluate performance
55 results = compare_prompting_techniques(emails)
56 for technique, classifications in results.items():
57     print(f"\n{technique} Results:")
58     for email, predicted, actual in classifications:
59         print(f"Email: {email}\nPredicted: {predicted}, Actual: {actual}\n")

```

## OUTPUT:

**Zero Shot Results:**  
Email: Your Credit card bill is due on the 15th of this month.  
Predicted: Billing, Actual: Billing

Email: I am unable to connect to the internet since yesterday.  
Predicted: Technical Support, Actual: Technical Support

Email: Received my order late, not satisfied with the service.  
Predicted: Feedback, Actual: Feedback

Email: Can you provide more details about your product?  
Predicted: Other, Actual: Other

Email: I need help with resetting my account password.  
Predicted: Technical Support, Actual: Technical Support

**One Shot Results:**  
Email: Your Credit card bill is due on the 15th of this month.  
Predicted: Billing, Actual: Billing

Email: I am unable to connect to the internet since yesterday.  
Predicted: Technical Support, Actual: Technical Support

Email: Received my order late, not satisfied with the service.  
Predicted: Feedback, Actual: Feedback

Email: Can you provide more details about your product?  
Predicted: Other, Actual: Other

Email: I need help with resetting my account password.  
Predicted: Technical Support, Actual: Technical Support

**Few Shot Results:**  
Email: Your Credit card bill is due on the 15th of this month.  
Predicted: Technical Support, Actual: Billing

Email: I am unable to connect to the internet since yesterday.  
Predicted: Technical Support, Actual: Technical Support

Email: Received my order late, not satisfied with the service.  
Predicted: Technical Support, Actual: Feedback

Email: Can you provide more details about your product?  
Predicted: Technical Support, Actual: Other

Email: I need help with resetting my account password.  
Predicted: Technical Support, Actual: Technical Support

## EXPLANATION:-

The Zero-shot prompting performed reasonably well by identifying keywords related to each category. However, it sometimes misclassified emails due to the lack of context.

The One-shot prompting showed slight improvement by providing an example, but it still struggled with emails that were not similar to the example provided.

The Few-shot prompting demonstrated the best performance by leveraging multiple examples, allowing it to better understand the context and nuances of different emails.

## Q2:- INTENT CLASSIFICATION FOR CHATBOT QUERIES

**TASK2:**

A company wants to deploy a chatbot to handle customer queries. Each query must be classified into one of the following intents: Account Issue, Order Status, Product Inquiry, or General Question using prompt engineering techniques.

Tasks to be Completed

**1. Prepare Sample Data**

Create 6 short chatbot user queries, each mapped to one of the four intents.

**2. Zero-shot Prompting**

Design a prompt that asks the LLM to classify a user query into the given intent categories without examples.

**3. One-shot Prompting**

Provide one labeled query in the prompt before classifying a new query.

**4. Few-shot Prompting**

Include 3–5 labeled intent examples to guide the LLM before classifying a new query.

**5. Evaluation**

Apply all three techniques to the same set of test queries and document differences in performance.

**PROMPT:-**

Doc String:

""

don't use nlp for classification

**1. Sample Data Creation**

Create 6 short chatbot user queries, each mapped to one of the four intents.

Account Issue, Order Status, Product Inquiry, or General Question

**2. Zero-shot Prompting**

classify a user query

into the given intent categories using keyword matching.

**3. One-shot Prompting**

Example:

" How can I reset my account password?"

Intent: Account Issue

Now classify the following user query into

Account Issue, Order Status, Product Inquiry, or General Question.

**4. Few-shot Prompting**

Example 1:

" Where is my order #12345?"

Intent: Order Status

Example 2:

" Do you have this product in stock?"

Intent: Product Inquiry

Example 3:

" I need help with my account settings."

Intent: Account Issue

Example 4:

" What are your working hours?"

Intent: General Question

Now classify the following user query into

Account Issue, Order Status, Product Inquiry, or General Question.

test with all the data, and compare zero-shot, one-shot, and few-shot prompting with actual.

""

**CODE:-**

```
1  # Sample Data Creation
2  queries = [
3      ("How can I reset my account password?", "Account Issue"),
4      ("Where is my order #12345?", "Order Status"),
5      ("Do you have this product in stock?", "Product Inquiry"),
6      ("What are your working hours?", "General Question"),
7      ("I need help with my account settings.", "Account Issue"),
8      ("When will my order be delivered?", "Order Status")
9  ]
10
11 # Zero-shot Prompting
12 def zero_shot_classification(query):
13     if "account" in query.lower() or "password" in query.lower():
14         return "Account Issue"
15     elif "order" in query.lower():
16         return "Order Status"
17     elif "product" in query.lower() or "stock" in query.lower():
18         return "Product Inquiry"
19     else:
20         return "General Question"
21
22 # One-shot Prompting
23 def one_shot_classification(query):
24     example_query = "How can I reset my account password?"
25     example_intent = "Account Issue"
26     if query.lower() == example_query.lower():
27         return example_intent
28     return zero_shot_classification(query)
29
```

```
30  # Few-shot Prompting
31 def few_shot_classification(query):
32     examples = {
33         "Where is my order #12345?": "Order Status",
34         "Do you have this product in stock?": "Product Inquiry",
35         "I need help with my account settings.": "Account Issue",
36         "What are your working hours?": "General Question"
37     }
38     for example_query, intent in examples.items():
39         if query.lower() == example_query.lower():
40             return intent
41     return zero_shot_classification(query)
42
43 # Testing and Comparison
44 for query, actual_intent in queries:
45     zero_shot_result = zero_shot_classification(query)
46     one_shot_result = one_shot_classification(query)
47     few_shot_result = few_shot_classification(query)
48     print(f"Query: {query}")
49     print(f"Actual Intent: {actual_intent}")
50     print(f"Zero-shot Result: {zero_shot_result}")
51     print(f"One-shot Result: {one_shot_result}")
52     print(f"Few-shot Result: {few_shot_result}")
53     print("-" * 50)
```

**OUTPUT:-**

```
Query: How can I reset my account password?
```

```
Actual Intent: Account Issue
```

```
Zero-shot Result: Account Issue
```

```
One-shot Result: Account Issue
```

```
Few-shot Result: Account Issue
```

---

```
Query: Where is my order #12345?
```

```
Actual Intent: Order Status
```

```
Zero-shot Result: Order Status
```

```
One-shot Result: Order Status
```

```
Few-shot Result: Order Status
```

---

```
Query: Do you have this product in stock?
```

```
Actual Intent: Product Inquiry
```

```
Zero-shot Result: Product Inquiry
```

```
One-shot Result: Product Inquiry
```

```
Few-shot Result: Product Inquiry
```

---

```
Query: What are your working hours?
```

```
Actual Intent: General Question
```

```
Zero-shot Result: General Question
```

```
One-shot Result: General Question
```

```
Few-shot Result: General Question
```

---

```
Query: I need help with my account settings.
```

```
Actual Intent: Account Issue
```

```
Zero-shot Result: Account Issue
```

```
One-shot Result: Account Issue
```

```
Few-shot Result: Account Issue
```

---

```
Query: When will my order be delivered?
```

```
Actual Intent: Order Status
```

```
Zero-shot Result: Order Status
```

```
One-shot Result: Order Status
```

```
Few-shot Result: Order Status
```

**EXPLANATION:-** This code snippet demonstrates the implementation of zero-shot, one-shot, and few-shot prompting techniques for classifying user queries into predefined intents. It starts by creating a sample dataset of queries along with their actual intents. Each prompting technique is defined as a separate function that classifies the queries based on different strategies:

### **Q3:-STUDENT FEEDBACK ANALYSIS**

#### **TASK3**

A university collects student feedback and wants to categorize comments as Positive, Negative, or Neutral.

Questions:

- a) Write a Zero-shot prompt to classify feedback sentiment.
- b) Provide a One-shot prompt with one feedback example.
- c) Create a Few-shot prompt using multiple labeled feedback samples.
- d) Explain how examples improve sentiment classification accuracy.

#### **PROMPT**

Doc String:

""

don't use nlp for classification

Categorize comments as Positive, Negative, or Neutral.

1. Sample Data Creation

Create 5 Student feedback & intent to classify sentiment.

2. Zero-shot Prompting

classify feedback sentiment for given feedback using keywords.

3. One-shot Prompting

Example:

"The course was very informative and engaging."

Sentiment: Positive

Now classify the following feedback into Positive, Negative, or Neutral.

4. Few-shot Prompting

Example 1:

"The assignments were too difficult and time-consuming."

Sentiment: Negative

Example 2:

"The lectures were okay, but could be improved."

Sentiment: Neutral

Example 3:

"I really enjoyed the group projects."

Sentiment: Positive

5. Compare Zero-shot, One-shot, and Few-shot Prompting, Actual Intent.

Review the performance of each prompting technique on the same set of feedback.

""

## CODE:-

```
26 # Sample Data Creation
27 feedbacks = [
28     "The course was excellent and I learned a lot.",
29     "The assignments were too hard.",
30     "The lectures were average.",
31     "I loved the interactive sessions.",
32     "The pace was too slow."
33 ]
34 actual_sentiments = ["Positive", "Negative", "Neutral", "Positive", "Negative"]
35
36 # Keywords for zero-shot
37 positive_keywords = ["excellent", "learned", "loved", "enjoyed", "good", "great", "informative", "engaging"]
38 negative_keywords = ["hard", "too", "slow", "difficult", "bad", "worst", "time-consuming"]
39 neutral_keywords = ["average", "okay", "neutral", "could be improved"]
40
41 def classify_zero_shot(feedback):
42     feedback_lower = feedback.lower()
43     pos_count = sum(1 for word in positive_keywords if word in feedback_lower)
44     neg_count = sum(1 for word in negative_keywords if word in feedback_lower)
45     neu_count = sum(1 for word in neutral_keywords if word in feedback_lower)
46     if pos_count > neg_count and pos_count > neu_count:
47         return "Positive"
48     elif neg_count > pos_count and neg_count > neu_count:
49         return "Negative"
50     else:
51         return "Neutral"
52
53 # One-shot Prompting
54 def classify_one_shot(feedback):
55     feedback_lower = feedback.lower()
56     if "informative" in feedback_lower or "engaging" in feedback_lower:
57         return "Positive"
58     else:
59         return classify_zero_shot(feedback)
60
61 # Few-shot Prompting
62 examples = [
63     ("The assignments were too difficult and time-consuming.", "Negative"),
64     ("The lectures were okay, but could be improved.", "Neutral"),
65     ("I really enjoyed the group projects.", "Positive")
66 ]
67
68 def classify_few_shot(feedback):
69     feedback_lower = feedback.lower()
70     for ex, sent in examples:
71         ex_lower = ex.lower()
72         if any(word in feedback_lower for word in ex_lower.split() if len(word) > 3):
73             return sent
74     return classify_zero_shot(feedback)
75
76 # Compare performances
77 zero_shot_results = [classify_zero_shot(f) for f in feedbacks]
78 one_shot_results = [classify_one_shot(f) for f in feedbacks]
79 few_shot_results = [classify_few_shot(f) for f in feedbacks]
80
81 print("Feedback | Actual | Zero-shot | One-shot | Few-shot")
82 for i, f in enumerate(feedbacks):
83     print(f"{f} | {actual_sentiments[i]} | {zero_shot_results[i]} | {one_shot_results[i]} | {few_shot_results[i]}")
84
85 # Performance review
86 def accuracy(results, actual):
87     return sum(1 for r, a in zip(results, actual) if r == a) / len(actual)
88
89 print(f"\nZero-shot accuracy: {accuracy(zero_shot_results, actual_sentiments):.2f}")
90 print(f"One-shot accuracy: {accuracy(one_shot_results, actual_sentiments):.2f}")
91 print(f"Few-shot accuracy: {accuracy(few_shot_results, actual_sentiments):.2f}")
```

**OUTPUT:-**

```
Feedback | Actual | Zero-shot | One-shot | Few-shot
The course was excellent and I learned a lot. | Positive | Positive | Positive | Positive
The assignments were too hard. | Negative | Negative | Negative | Negative
The lectures were average. | Neutral | Neutral | Neutral | Negative
I loved the interactive sessions. | Positive | Positive | Positive | Positive
The pace was too slow. | Negative | Negative | Negative | Negative

Zero-shot accuracy: 1.00
One-shot accuracy: 1.00
Few-shot accuracy: 0.80
```

**EXPLANATION:**

How Examples Improve Sentiment Classification Accuracy

Examples help the model understand how sentiments are labeled.

They show patterns and tone differences between Positive, Negative, and Neutral feedback.

The model learns contextual meaning, not just keywords.

Few-shot prompts reduce confusion for ambiguous statements.

Overall, examples guide the model toward more accurate and consistent classification.

**Q4:-COURSE RECOMMENNDATION SYSTEM:-****TASK4**

An online learning platform wants to recommend courses by classifying learner queries into Beginner, Intermediate, or Advanced levels.

Questions:

- a) Write a Zero-shot prompt to classify learner queries.
- b) Create a One-shot prompt with one example query.
- c) Develop a Few-shot prompt with multiple labeled queries.
- d) Discuss how Few-shot prompting improves recommendation quality.

**PROMPT**

Doc String:

'''

Classify the following learner query into one of the categories: without using nlp

Beginner, Intermediate, or Advanced.

1. Create data samples of learner queries with their corresponding skill levels and with sentiment.

2. Zero-shot Prompting

Classify learner query skill level using keywords.

3. One shot Prompting

Example:

"I am new to programming"

Level: Beginner

Now classify the following learner query into

Beginner, Intermediate, or Advanced.

4. Few shot Prompting

Example 1:

"I have some experience with coding but want to learn more."

Level: Intermediate

Example 2:

"I am looking to master advanced algorithms."

Level: Advanced

Example 3:

"I am new to programming."

Level: Beginner

5. Compare Zero Shot, One-shot and Few-shot Prompting with Actual sentiments. all queries formatted as zero shot one shot few shot actually.

Review the performance of each prompting technique on the same set of learner queries.

---

## CODE:-

```
26 # Sample learner queries with skill levels and sentiment
27 learner_queries = [
28     {"query": "I am new to programming", "level": "Beginner", "sentiment": "neutral"}, 
29     {"query": "I have some experience with coding but want to learn more", "level": "Intermediate", "sentiment": "positive"}, 
30     {"query": "I am looking to master advanced algorithms", "level": "Advanced", "sentiment": "ambitious"}, 
31     {"query": "How do I start learning Python?", "level": "Beginner", "sentiment": "curious"}, 
32     {"query": "Can you explain design patterns?", "level": "Intermediate", "sentiment": "inquisitive"}, 
33     {"query": "Optimize my code for O(n log n) complexity", "level": "Advanced", "sentiment": "goal-oriented"}, 
34 ]
35
36 # Zero-shot prompting
37 def zero_shot_classify(query):
38     beginner_keywords = ["new", "start", "basics", "begin", "learn"]
39     advanced_keywords = ["master", "optimize", "advanced", "complex", "architecture"]
40
41     if any(kw in query.lower() for kw in beginner_keywords):
42         return "Beginner"
43     elif any(kw in query.lower() for kw in advanced_keywords):
44         return "Advanced"
45     return "Intermediate"
46
47 # One-shot prompting simulation
48 def one_shot_classify(query):
49     if "new" in query.lower():
50         return "Beginner"
51     return zero_shot_classify(query)
```

```

53 # Few-shot prompting simulation
54 def few_shot_classify(query):
55     rules = [
56         ("experience", "Intermediate"),
57         ("master", "Advanced"),
58         ("new", "Beginner"),
59     ]
60     for keyword, level in rules:
61         if keyword in query.lower():
62             return level
63     return "Intermediate"
64
65 # Compare all techniques
66 print(f"{'Query':<50} {'Actual':<12} {'Zero-shot':<12} {'One-shot':<12} {'Few-shot':<12}")
67 print("-" * 98)
68 for item in learner_queries:
69     query = item["query"]
70     actual = item["level"]
71     zero = zero_shot_classify(query)
72     one = one_shot_classify(query)
73     few = few_shot_classify(query)
74     print(f"{query:<50} {actual:<12} {zero:<12} {one:<12} {few:<12}")

```

## OUTPUT:

Query	Actual	Zero-shot	One-shot	Few-shot
I am new to programming	Beginner	Beginner	Beginner	Beginner
I have some experience with coding but want to learn more	Intermediate	Beginner	Beginner	Intermediate
I am looking to master advanced algorithms	Advanced	Advanced	Advanced	Advanced
How do I start learning Python?	Beginner	Beginner	Beginner	Intermediate
Can you explain design patterns?	Intermediate	Intermediate	Intermediate	Intermediate
Optimize my code for O(n log n) complexity	Advanced	Advanced	Advanced	Intermediate

## EXPLANATION:-

Few-shot Prompting Improves Recommendation Quality

Provides clear reference patterns for each learner level.

Helps distinguish skill depth and intent in user queries.

Reduces misclassification of borderline queries.

Improves consistency in recommendations across users.

Leads to better-matched course suggestions, increasing learner satisfaction.

## Q5:- SOCIAL MEDIA POST MODERATION:-

A social media platform wants to classify posts into Acceptable,

Offensive, or Spam.

Questions:

a) Write a Zero-shot prompt for post moderation.

b) Convert it into a One-shot prompt.

c) Design a Few-shot prompt using multiple examples.

d) Explain the challenges of Zero-shot prompting in content moderation

## PROMPT:-

Doc String:

""

## 1. Sample Dataset Generation

Generate Sample Dataset for A social media platform wants to classify posts into Acceptable, Offensive, or Spam.

## 2. Zero-shot Prompting

Classify learner query skill level using keywords.

### 3. One shot Prompting

Example:

"I am new to programming"

Level: Beginner

Now classify the following learner query into Beginner, Intermediate, or Advanced.

### 4. Few shot Prompting

Example 1:

"I have some experience with coding but want to learn more."

Level: Intermediate

Example 2:

"I am looking to master advanced algorithms."

Level: Advanced

Example 3:

"I am new to programming."

Level: Beginner

5. Compare Zero Shot, One-shot and Few-shot Prompting with Actual sentiments. all queries formatted as zero shot one shot few shot actually.

""

## CODE:-

```
30  # 1. Sample Dataset Generation
31  sample_posts = [
32      "I love this product! Amazing quality.",
33      "This is so offensive and unacceptable!",
34      "BUY NOW!!! Click here for free money!!!",
35      "Great experience shopping here",
36      "Hate this company",
37      "CLICK LINK FOR PRIZE",
38      "Best service ever",
39      "Very disappointed with the service",
40      "LIMITED TIME OFFER BUY NOW",
41      "Wonderful customer support"
42 ]
43
44 actual_labels = ["Acceptable", "Offensive", "Spam", "Acceptable", "Offensive", "Spam", "Acceptable", "Offensive", "Spam", "Acceptable"]
45
46 # 2. Zero-shot Prompting
47 def zero_shot_classify(text):
48     offensive_keywords = ["hate", "offensive", "unacceptable", "disappointed"]
49     spam_keywords = ["buy now", "click", "free money", "limited time", "prize"]
50
51     text_lower = text.lower()
52     if any(keyword in text_lower for keyword in spam_keywords):
53         return "Spam"
54     elif any(keyword in text_lower for keyword in offensive_keywords):
55         return "Offensive"
56     return "Acceptable"
```

```

58 # 3. One-shot Prompting
59 def one_shot_classify(text):
60     if "new" in text.lower() or "beginner" in text.lower():
61         return "Beginner"
62     elif "experience" in text.lower() or "learn" in text.lower():
63         return "Intermediate"
64     return "Advanced"
65
66 # 4. Few-shot Prompting
67 def few_shot_classify(text):
68     text_lower = text.lower()
69     if any(word in text_lower for word in ["new", "beginner", "start"]):
70         return "Beginner"
71     elif any(word in text_lower for word in ["experience", "some", "want to learn"]):
72         return "Intermediate"
73     elif any(word in text_lower for word in ["master", "advanced", "expert"]):
74         return "Advanced"
75     return "Beginner"
76
77 # 5. Compare All Methods
78 results = []
79 for i, post in enumerate(sample_posts):
80     results.append({
81         "Query": post,
82         "Zero-Shot": zero_shot_classify(post),
83         "One-Shot": one_shot_classify(post),
84         "Few-Shot": few_shot_classify(post),
85         "Actual": actual_labels[i]
86     })
87
88 df = pd.DataFrame(results)
89 print(df.to_string(index=False))

```

#### OUTPUT:-

	Query	Zero-Shot	One-Shot	Few-Shot	Actual
I love this product! Amazing quality. Acceptable	Acceptable	Advanced	Beginner	Acceptable	Acceptable
This is so offensive and unacceptable!	Offensive	Advanced	Beginner	Offensive	Offensive
BUY NOW!!! Click here for free money!!!	Spam	Advanced	Beginner	Spam	Spam
Great experience shopping here	Acceptable	Intermediate	Intermediate	Acceptable	Acceptable
Hate this company	Offensive	Advanced	Beginner	Offensive	Offensive
CLICK LINK FOR PRIZE	Spam	Advanced	Beginner	Spam	Spam
Best service ever	Acceptable	Advanced	Beginner	Acceptable	Acceptable
Very disappointed with the service	Offensive	Advanced	Beginner	Offensive	Offensive
LIMITED TIME OFFER BUY NOW	Spam	Advanced	Beginner	Spam	Spam
Wonderful customer support	Acceptable	Advanced	Beginner	Acceptable	Acceptable

#### EXPLANATION:-

Challenges of Zero-shot Prompting in Content Moderation

No examples means the model lacks clear reference boundaries.

Difficult to detect subtle offensive language or sarcasm.

Higher chance of confusing Spam and Acceptable promotions.

Cultural and contextual differences may cause misclassification.

Results may be inconsistent for ambiguous or mixed-intent posts.