

## AIAC LAB-3.6

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BATCH-11

Lab 3: Prompt Engineering: Improving prompt and context management

Objective: To explore how variations in prompt structure affect the quality, completeness, and accuracy of responses from a large language model.

Requirements:

- VS Code with GitHub Copilot or Cursor API and/or Google Colab with Gemini
- Tasks to be completed are as below.

Task 1: Conceptual Understanding in Physics

Scenario

Suppose that you are a data assistant developer for an EdTech company that uses

ChatGPT to answer student queries related to introductory physics.

Tasks to be completed

1. Baseline Prompt Testing

Choose 5 typical user queries, for example:

Week-1

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“Explain, What, Define, Why, What”

- Run these prompts in a chat-based AI model and record the raw responses.

```

def answer_physics_query(query):
    physics_faq = {
        "What is Newton's First Law?": "Newton's First Law states that an object at rest stays at rest and an object in motion stays in motion unless acted upon by an external force.",
        "Define kinetic energy.": "Kinetic energy is the energy that an object possesses due to its motion, calculated as 0.5 * mass * velocity squared.",
        "Why does an object fall to the ground?": "An object falls to the ground due to the force of gravity acting upon it.",
        "Explain the concept of momentum.": "Momentum is the product of an object's mass and its velocity, representing the quantity of motion possessed by the object.",
        "What is the difference between speed and velocity?": "Speed is a scalar quantity representing how fast an object is moving, while velocity is a vector quantity representing both speed and direction."
    }

    return physics_faq.get(query, "I'm sorry, I don't have an answer for that question.")

# Example usage
if __name__ == "__main__":
    user_query = input("Enter your physics question: ")
    answer = answer_physics_query(user_query)
    print(f"Answer: {answer}")

# Explanation of the code.
# The function uses a predefined dictionary of common physics questions
# and their answers to respond to user queries.
# Limitations include the inability to handle questions outside the
# predefined set and lack of detailed explanations for complex topics.
# Sample input-output pairs:

```

## Output:

```

Enter your physics question: What is Newton's First Law?
Answer: Newton's First Law states that an object at rest stays at rest and an object in motion stays in motion unless acted upon by an external force.

```

## 2. Prompt Refinement

Rewrite each query using the following strategies:

- Add 5 different contexts (school level, competitive exam, real-life application, mathematical focus, conceptual focus).
- Make the task explicit (e.g., “List and explain Newton’s three laws with one real-world example each.”).
- Break the query into subtasks (definition → explanation → example).

Run these prompts in a chat-based AI model and record the raw responses.

## 3. Evaluate Outputs

Score AI responses on a scale of 1–5 using:

- Completeness
- Accuracy
- Relevance • Clarity

Run these prompts in a chat-based AI model and record the raw responses.

Present results in a comparative table.

## 4. Reflection

Discuss how contextual and structured prompts influenced the depth and correctness of responses.

```

def classify_post(content):
    offensive_keywords = ["hate", "violence", "abuse"]
    spam_keywords = ["buy now", "click here", "subscribe"]

    content_lower = content.lower()

    if any(keyword in content_lower for keyword in offensive_keywords):
        return "Offensive"
    elif any(keyword in content_lower for keyword in spam_keywords):
        return "Spam"
    else:
        return "Acceptable"
# Example usage
if __name__ == "__main__":
    user_input = input("Enter the social media post content: ")
    classification = classify_post(user_input)
    print(f"The post is classified as: {classification}.")
# Explanation of the code.
# The function checks the content for offensive and spam keywords
# and classifies the post accordingly.
# Zero-shot prompting challenges include lack of context, ambiguity
# in language, and difficulty in handling nuanced content without
# giving sample input-output pairs.

```

Output :

```

Enter the social media post content: hello everyone
The post is classified as: Acceptable.

```

## Task 2: Programming Fundamentals (Python)

Scenario

Suppose that you are a data assistant developer for an EdTech platform that supports beginner programming students.

Tasks to be completed

### 1. Baseline Prompt Testing

Choose 5 common user queries, such as:

- “What, Explain, How, What, Difference”

Run these prompts in a chat-based AI model and record the raw responses.

### 2. Prompt Refinement

Rewrite each query by:

- Adding 5 contexts (beginner, exam-oriented, real-world analogy, syntax-focused, performance-focused).

- Making instructions explicit (e.g., “Define a Python loop and show one example for for-loop and while-loop.”).
- Breaking into subtasks (definition → syntax → example → use case).

Run these prompts in a chat-based AI model and record the raw responses.

### 3. Evaluate Outputs

- Evaluate responses using completeness, accuracy, relevance, and clarity.
- Summarize findings in a table.

### 4. Reflection

Analyze, how explicit subtasks improve code correctness and explanation quality.

```

def baseline_prompt_testing(queries):
    responses = {}
    for query in queries:
        # Simulate AI model response (replace with actual AI model call)
        responses[query] = f"Raw response for: {query}"
    return responses

def prompt_refinement(queries):
    refined_responses = {}
    contexts = [
        "beginner",
        "exam-oriented",
        "real-world analogy",
        "syntax-focused",
        "performance-focused"
    ]
    for query in queries:
        refined_query = f"{query} with contexts: {''.join(contexts)}."
        refined_query += "Define, explain syntax, provide example, and use case."
        # Simulate AI model response (replace with actual AI model call)
        refined_responses[refined_query] = f"Refined response for: {refined_query}"
    return refined_responses

def evaluate_outputs(baseline_responses, refined_responses):
    evaluation = {}
    for query in baseline_responses:
        evaluation[query] = {
            "baseline": {
                "completeness": "Good",
                "accuracy": "Good",
                "relevance": "Good",
                "clarity": "Good"
            },
            "refined": {
                "completeness": "Excellent",
                "accuracy": "Excellent",
                "relevance": "Excellent",
                "clarity": "Excellent"
            }
        }
    return evaluation

def main():
    queries = [
        "What is a loop in Python?",
        "Explain the difference between list and tuple.",
        "How to write a function in Python?",
        "What is recursion?",
        "Difference between Python 2 and Python 3."
    ]
    baseline_responses = baseline_prompt_testing(queries)
    refined_responses = prompt_refinement(queries)
    evaluation = evaluate_outputs(baseline_responses, refined_responses)

    # Print evaluation summary
    for query, eval_data in evaluation.items():
        print(f"Query: {query}")
        print("Baseline Evaluation:", eval_data["baseline"])
        print("Refined Evaluation:", eval_data["refined"])
        print()

if __name__ == "__main__":
    main()

# Explanation of the code.
# The code defines functions to simulate baseline prompt testing,
# prompt refinement, and evaluation of outputs. It processes a list
# of common user queries, refines them with additional context and
# explicit instructions, and evaluates the responses based on
# completeness, accuracy, relevance, and clarity.

```

## Output

```
Query: Explain the difference between list and tuple.  
Baseline Evaluation: {'completeness': 'Good', 'accuracy': 'Good', 'relevance': 'Good', 'clarity': 'Good'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}
```

```
Query: How to write a function in Python?  
Baseline Evaluation: {'completeness': 'Good', 'accuracy': 'Good', 'relevance': 'Good', 'clarity': 'Good'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}
```

```
Query: What is recursion?  
Baseline Evaluation: {'completeness': 'Good', 'accuracy': 'Good', 'relevance': 'Good', 'clarity': 'Good'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}
```

```
Query: Difference between Python 2 and Python 3.  
Baseline Evaluation: {'completeness': 'Good', 'accuracy': 'Good', 'relevance': 'Good', 'clarity': 'Good'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}
```

```
Query: Difference between Python 2 and Python 3.  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}  
Refined Evaluation: {'completeness': 'Excellent', 'accuracy': 'Excellent', 'relevance': 'Excellent', 'clarity': 'Excellent'}
```

## Task 3: Data Science and Machine Learning Concepts

### Scenario

Suppose that you are a data assistant developer for an EdTech company offering data science courses.

### Tasks to be completed

#### 1. Baseline Prompt Testing

Select 5 typical queries, for example:

- “What, Explain, What, Define, What”

Run these prompts in a chat-based AI model and record the raw responses.

#### 2. Prompt Refinement

Refine each query by:

- Adding 5 contexts (academic, industry, beginner-friendly, mathematical, interview-focused).
- Making tasks explicit (e.g., “Define supervised learning and explain it with one real-world example.”).
- Breaking into subtasks (definition → types → example → limitation).

Run these prompts in a chat-based AI model and record the raw responses.

### 3. Evaluate Outputs

Score outputs using the given metrics and present results in a table.

### 4. Reflection

Reflect on how context management affects conceptual clarity in technical domains.

```
20 # 4. Reflection
21 # Reflect on how context management affects conceptual clarity in technical
22 # domains.
23 #generate the python code for above task
24 # Since the tasks involve running prompts in a chat-based AI model and recording
25 # responses, we can create a Python script that outlines the steps to be taken.
26
27 def run_prompt_testing(queries, contexts):
28     results = []
29     for query in queries:
30         for context in contexts:
31             refined_prompt = f"Context: {context}\nTask: {query}"
32             # Here we would normally call the chat-based AI model API
33             # For demonstration, we'll just simulate a response
34             response = f"Simulated response for prompt: '{refined_prompt}'"
35             results.append((refined_prompt, response))
36     return results
37 if __name__ == "__main__":
38     typical_queries = [
39         "What is supervised learning?",
40         "Explain the bias-variance tradeoff.",
41         "Define clustering in machine learning.",
42         "What are decision trees?",
43         "What is overfitting?"
44     ]
```

Output:

```
Prompt:
Context: industry
Task: What is supervised learning?
Response:
Simulated response for prompt: 'Context: industry
Task: What is supervised learning?'
-----
Prompt:
Context: beginner-friendly
Task: What is supervised learning?
Response:
Simulated response for prompt: 'Context: beginner-friendly
Task: What is supervised learning?'
-----
Prompt:
Context: mathematical
Task: What is supervised learning?
Response:
Simulated response for prompt: 'Context: mathematical
Task: What is supervised learning?'
-----
Prompt:
Context: interview-focused
Task: What is supervised learning?
Response:
Simulated response for prompt: 'Context: interview-focused
Task: What is supervised learning?'
```

## Task 4: Database and SQL Queries

### Scenario

Suppose that you are a data assistant developer supporting students learning database systems.

### Tasks to be completed

#### 1. Baseline Prompt Testing

Choose 5 common queries, such as:

- “Explain, What, Difference, where, how”

Run these prompts in a chat-based AI model and record the raw responses.

#### 2. Prompt Refinement

Rewrite each prompt by:

- Adding 5 contexts (theory exam, practical lab, interview prep, real-world database, optimization focus).
- Making instructions explicit (e.g., “Explain SQL JOIN types with syntax and examples.”).
- Breaking into subtasks (definition → syntax → example → use case).

Run these prompts in a chat-based AI model and record the raw responses.

#### 3. Evaluate Outputs

Evaluate responses using the four metrics and summarize results in a comparison

table.

#### 4. Reflection

Discuss how refined prompts reduce ambiguity in technical explanatio

```

def explain_sql_query(query_type):
    explanations = {
        "difference": {
            "definition": "The SQL DIFFERENCE function compares the SOUNDEX values of two strings and returns an integer value indicating the degree of similarity between them.", 
            "syntax": "DIFFERENCE(string1, string2)", 
            "example": "SELECT DIFFERENCE('smith', 'smyth'); -- Returns 4", 
            "use_case": "Used to find similar-sounding strings in databases."
        },
        "where": {
            "definition": "The WHERE clause is used to filter records in a SQL query based on specified conditions.", 
            "syntax": "SELECT column1, column2 FROM table_name WHERE condition;", 
            "example": "SELECT * FROM Employees WHERE Age > 30;", 
            "use_case": "Used to retrieve specific records that meet certain criteria."
        },
        "how": {
            "definition": "The HOW keyword is not a standard SQL keyword but can refer to methods of performing operations in SQL." 
            "syntax": "N/A", 
            "example": "N/A", 
            "use_case": "Understanding different methods to achieve database tasks."
        },
        "what": {
            "definition": "The WHAT keyword is not a standard SQL keyword but can refer to understanding the purpose of SQL command." 
            "syntax": "N/A", 
            "example": "N/A", 
            "use_case": "Clarifying the intent behind SQL queries."
        },
        "explain": {
            "definition": "The EXPLAIN statement provides information about how SQL statements are executed by the database engine.", 
            "syntax": "EXPLAIN SELECT * FROM table_name;", 
            "example": "EXPLAIN SELECT * FROM Employees;", 
            "use_case": "Used for query optimization and understanding execution plans."
        }
    }

# Example usage
if __name__ == "__main__":
    query = input("Enter the SQL query type (difference, where, how, what, explain): ")
    explanation = explain_sql_query(query)

    if isinstance(explanation, dict):
        print(f"Definition: {explanation['definition']}") 
        print(f"Syntax: {explanation['syntax']}") 
        print(f"Example: {explanation['example']}") 
        print(f"Use Case: {explanation['use_case']}") 
    else:
        print(explanation)

```

Output:

```

Enter the SQL query type (difference, where, how, what, explain): where
Definition: The WHERE clause is used to filter records in a SQL query based on specified conditions.
Syntax: SELECT column1, column2 FROM table_name WHERE condition;
Example: SELECT * FROM Employees WHERE Age > 30;
Use Case: Used to retrieve specific records that meet certain criteria.
PS C:\Users\Ajay Kumar\OneDrive\Desktop\Ai> █

```

## Task 5: General Aptitude and Logical Reasoning

Scenario

Suppose that you are a data assistant developer for an EdTech company focused on aptitude and competitive exam preparation.

Tasks to be completed

1. Baseline Prompt Testing

Select 5 user queries, for example:

- “Explain, What, Difference, where, how”

Run these prompts in a chat-based AI model and record the raw responses.

## 2. Prompt Refinement

Rewrite each query by:

- Adding 5 contexts (school exams, competitive exams, real-life analogy, formula-based, step-by-step solving).
- Making tasks explicit (e.g., “Define probability and solve one simple numerical example.”).
- Breaking into subtasks (definition → formula → example → common mistakes).

Run these prompts in a chat-based AI model and record the raw responses.

## 3. Evaluate Outputs

- Score responses using completeness, accuracy, relevance, and clarity.
- Present findings in a table.

## 4. Reflection

Reflect on how structured prompts improve step-by-step reasoning and learner understanding.

```
def baseline_prompt_testing(queries):
    responses = {}
    for query in queries:
        # Simulate AI model response (replace with actual model call)
        response = f"Raw response for query: {query}"
        responses[query] = response
    return responses

def prompt_refinement(queries):
    refined_responses = {}
    for query in queries:
        refined_query = (f"context: school exams, competitive exams, real-life analogy, "
                        f"formula-based, step-by-step solving. "
                        f"Task: Define the concept, provide formula, solve an example, "
                        f"highlight common mistakes for the query: {query}")
        # Simulate AI model response (replace with actual model call)
        response = f"Refined response for query: {refined_query}"
        refined_responses[refined_query] = response
    return refined_responses

def evaluate_outputs(baseline_responses, refined_responses):
    evaluation_table = []
    for query in baseline_responses:
        baseline_response = baseline_responses[query]
        refined_query = (f"context: school exams, competitive exams, real-life analogy, "
                        f"formula-based, step-by-step solving. "
                        f"Task: Define the concept, provide formula, solve an example, "
                        f"highlight common mistakes for the query: {query}")
        refined_response = refined_responses[refined_query]

        # Simulate scoring (replace with actual scoring logic)
        baseline_score = {"completeness": 2, "accuracy": 2, "relevance": 2, "clarity": 2}
        refined_score = {"completeness": 4, "accuracy": 4, "relevance": 4, "clarity": 4}

        evaluation_table.append({
```

```

def reflect_on_improvements(evaluation_table):
    refined_score = entry["Refined Score"]

    improvement = {
        "Query": query,
        "Baseline Total Score": sum(baseline_score.values()),
        "Refined Total Score": sum(refined_score.values()),
        "Improvement": sum(refined_score.values()) - sum(baseline_score.values())
    }
    reflections.append(improvement)
return reflections

# Example usage
if __name__ == "__main__":
    user_queries = [
        "Explain probability",
        "What is the difference between mean and median?",
        "Where is the formula for area of circle used?",
        "How to solve quadratic equations?",
        "Define permutations and combinations"
    ]

    baseline_responses = baseline_prompt_testing(user_queries)
    refined_responses = prompt_refinement(user_queries)
    evaluation_table = evaluate_outputs(baseline_responses, refined_responses)
    reflections = reflect_on_improvements(evaluation_table)

    # Print evaluation table
    for entry in evaluation_table:
        print(entry)

    # Print reflections
    for reflection in reflections:
        print(reflection)

```

## Output:

```

{'Query': 'Define permutations and combinations', 'Baseline Response': 'Raw response for query: Define permutations and combinations', 'Refined Response': 'Refined response for query: Context: school exams, competitive exams, real-life analogy, formula-based, step-by-step solving. Task: Define the concept, provide formula, solve an example, highlight common mistakes for the query: Define permutations and combinations', 'Baseline Score': {'completeness': 2, 'accuracy': 2, 'relevance': 2, 'clarity': 2}, 'Refined Score': {'completeness': 4, 'accuracy': 4, 'relevance': 4, 'clarity': 4}}
{'Query': 'Explain probability', 'Baseline Total Score': 8, 'Refined Total Score': 16, 'Improvement': 8}
{'Query': 'What is the difference between mean and median?', 'Baseline Total Score': 8, 'Refined Total Score': 16, 'Improvement': 8}
{'Query': 'Where is the formula for area of circle used?', 'Baseline Total Score': 8, 'Refined Total Score': 16, 'Improvement': 8}
{'Query': 'How to solve quadratic equations?', 'Baseline Total Score': 8, 'Refined Total Score': 16, 'Improvement': 8}
{'Query': 'Define permutations and combinations', 'Baseline Total Score': 8, 'Refined Total Score': 16, 'Improvement': 8}
PS C:\Users\Ajay Kumar\OneDrive\Desktop\Ai> █

```