

ASSIGNMENT-02

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Batch:06

Q) Task 1: Word Frequency from Text File

❖ Scenario:

You are analyzing log files for keyword frequency.

❖ Task:

Use Gemini to generate Python code that reads a text file and counts word frequency, then explains the code.

❖ Expected Output:

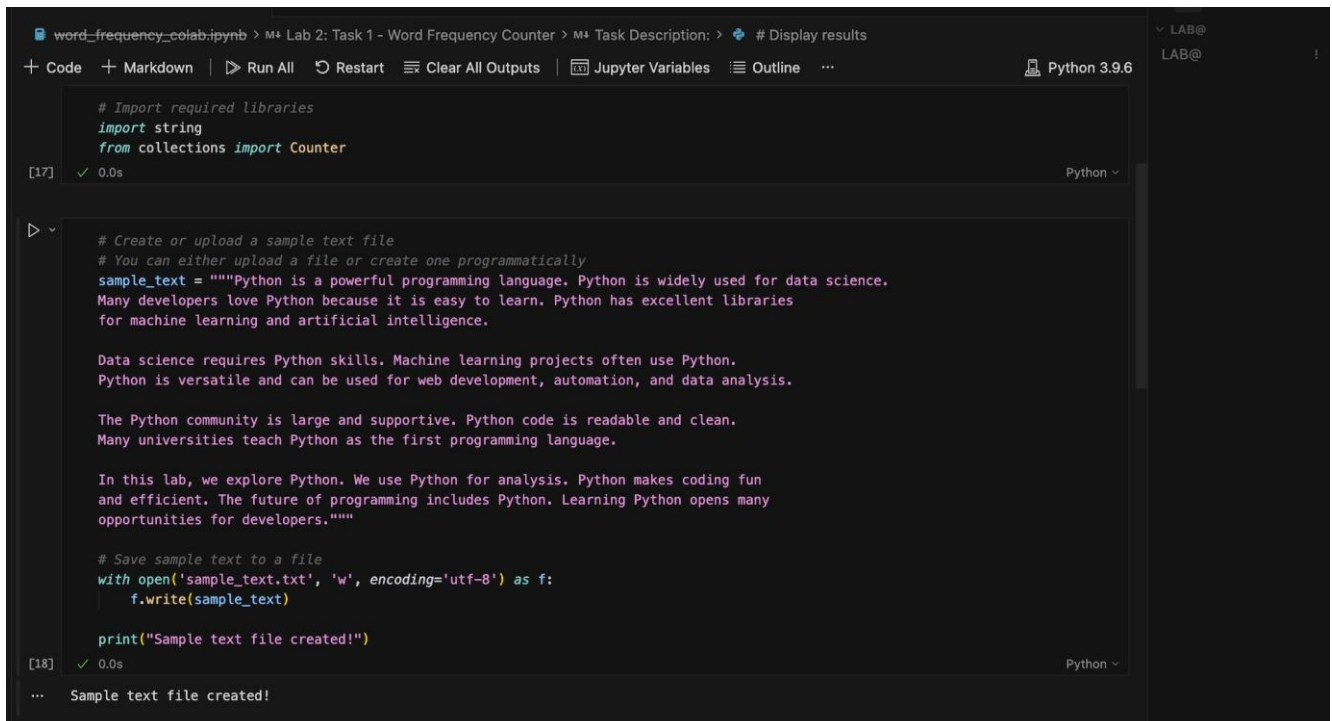
- Working code
- Explanation ➤ Screenshot

Solution:

PROMPT

Generate a Python program in Google Colab that reads a text file and counts the frequency of each word.

CODE:



```
word_frequency_colab.ipynb > M4 Lab 2: Task 1 - Word Frequency Counter > M4 Task Description: > # Display results
+ Code + Markdown | ▶ Run All ⏮ Restart ⌵ Clear All Outputs | 📄 Jupyter Variables 📄 Outline ... Python 3.9.6

# Import required libraries
import string
from collections import Counter

[17] ✓ 0.0s Python ▾

▶ ▾
# Create or upload a sample text file
# You can either upload a file or create one programmatically
sample_text = """Python is a powerful programming language. Python is widely used for data science.
Many developers love Python because it is easy to learn. Python has excellent libraries
for machine learning and artificial intelligence.

Data science requires Python skills. Machine learning projects often use Python.
Python is versatile and can be used for web development, automation, and data analysis.

The Python community is large and supportive. Python code is readable and clean.
Many universities teach Python as the first programming language.

In this lab, we explore Python. We use Python for analysis. Python makes coding fun
and efficient. The future of programming includes Python. Learning Python opens many
opportunities for developers."""

# Save sample text to a file
with open('sample_text.txt', 'w', encoding='utf-8') as f:
    f.write(sample_text)

print("Sample text file created!")

[18] ✓ 0.0s Python ▾

... Sample text file created!
```

```
word_frequency_colab.ipynb > M* Lab 2: Task 1 - Word Frequency Counter > M* Task Description: > # Display results
+ Code + Markdown | ▶ Run All ⏮ Restart ⏹ Clear All Outputs | 📄 Jupyter Variables 📖 Outline ... Python 3.9.6
LAB@
LAB@

def count_word_frequency(filename):
    """
    Read a text file and count the frequency of each word.

    Args:
        filename (str): Path to the text file to analyze

    Returns:
        Counter: Counter object with words as keys and frequencies as values
    """
    try:
        # Open and read the file
        with open(filename, 'r', encoding='utf-8') as file:
            text = file.read()

        # Convert to lowercase and remove punctuation
        translator = str.maketrans('', '', string.punctuation)
        text = text.translate(translator).lower()

        # Split text into words
        words = text.split()

        # Count word frequencies using Counter
        word_freq = Counter[str](words)

        return word_freq

    except FileNotFoundError:
        print(f"Error: File '{filename}' not found.")
        return None
    except Exception as e:
        print(f"Error reading file: {e}")
        return None

[19] ✓ 0.0s Python ▾

# Execute the word frequency analysis
filename = 'sample_text.txt'
word_freq = count_word_frequency(filename)

[20] ✓ 0.0s Python ▾
```

```
# Display results
if word_freq:
    print("\n" + "="*50)
    print("WORD FREQUENCY ANALYSIS")
    print("="*50)

    # Display top 20 most common words
    print("\nTop 20 Most Frequent Words:")
    print("-"*50)
    print(f"{'Word':<20} {'Frequency':<15} {'Percentage':<15}")
    print("-"*50)

    total_words = sum(word_freq.values())

    for word, count in word_freq.most_common(20):
        percentage = (count / total_words) * 100
        print(f"{'word':<20} {'count':<15} {'percentage':.2f}%")

    print("-"*50)
    print(f"\nTotal unique words: {len(word_freq)}")
    print(f"Total words: {total_words}")
    print("="*50)
```

OUTPUT:

```
=====
WORD FREQUENCY ANALYSIS
=====
```

```
Top 20 Most Frequent Words:
```

Word	Frequency	Percentage
python	15	13.64%
is	6	5.45%
and	6	5.45%
for	5	4.55%
programming	3	2.73%
data	3	2.73%
many	3	2.73%
learning	3	2.73%
the	3	2.73%
language	2	1.82%
used	2	1.82%
science	2	1.82%
developers	2	1.82%
machine	2	1.82%
use	2	1.82%
analysis	2	1.82%
...		

```
Total unique words: 64
```

```
Total words: 110
```

```
=====
Output is truncated. View as a scrollable element or open in a text editor. Adjust cell output settings...
```

CODE Explanation:

This Python program works by first importing the required modules to handle punctuation removal and word counting. The text file is opened in read mode and its content is read completely. Then, all punctuation marks are removed and the text is converted to lowercase so that words are counted correctly without case differences. After that, the text is split into individual words. The Counter function is used to count the number of times each word appears in the file. The program also includes error handling to display a message if the file is not found or if any other error occurs. Finally, the word frequencies are displayed in an organized format, making the output easy to understand

Q) Task 2: File Operations Using Cursor AI ❖

Scenario:

You are automating basic file operations.

❖ Task:

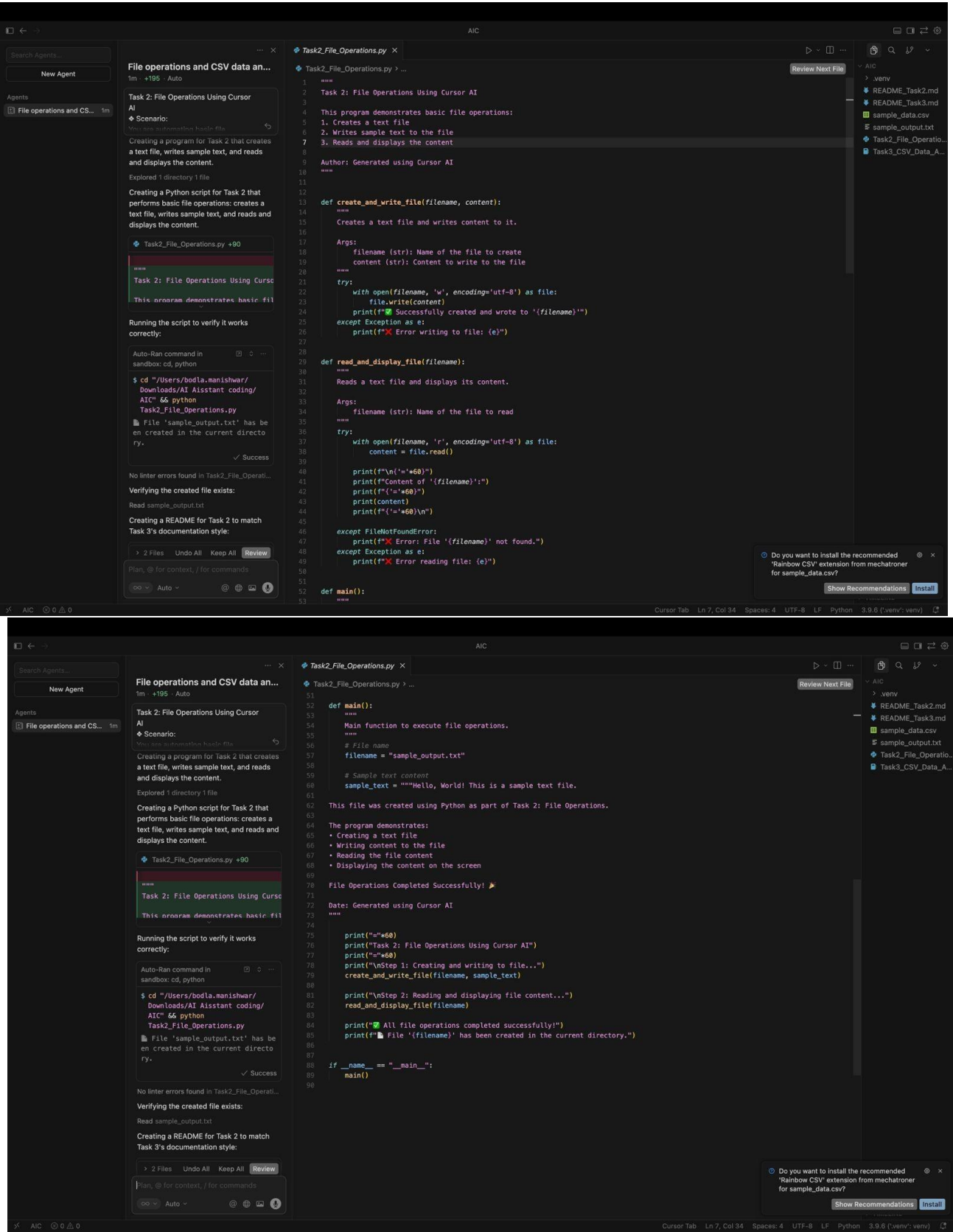
Use Cursor AI to generate a program that:

- Creates a text file
- Writes sample text
- Reads and displays the content ❖ Expected Output:
- Functional code ➤ Cursor AI screenshots

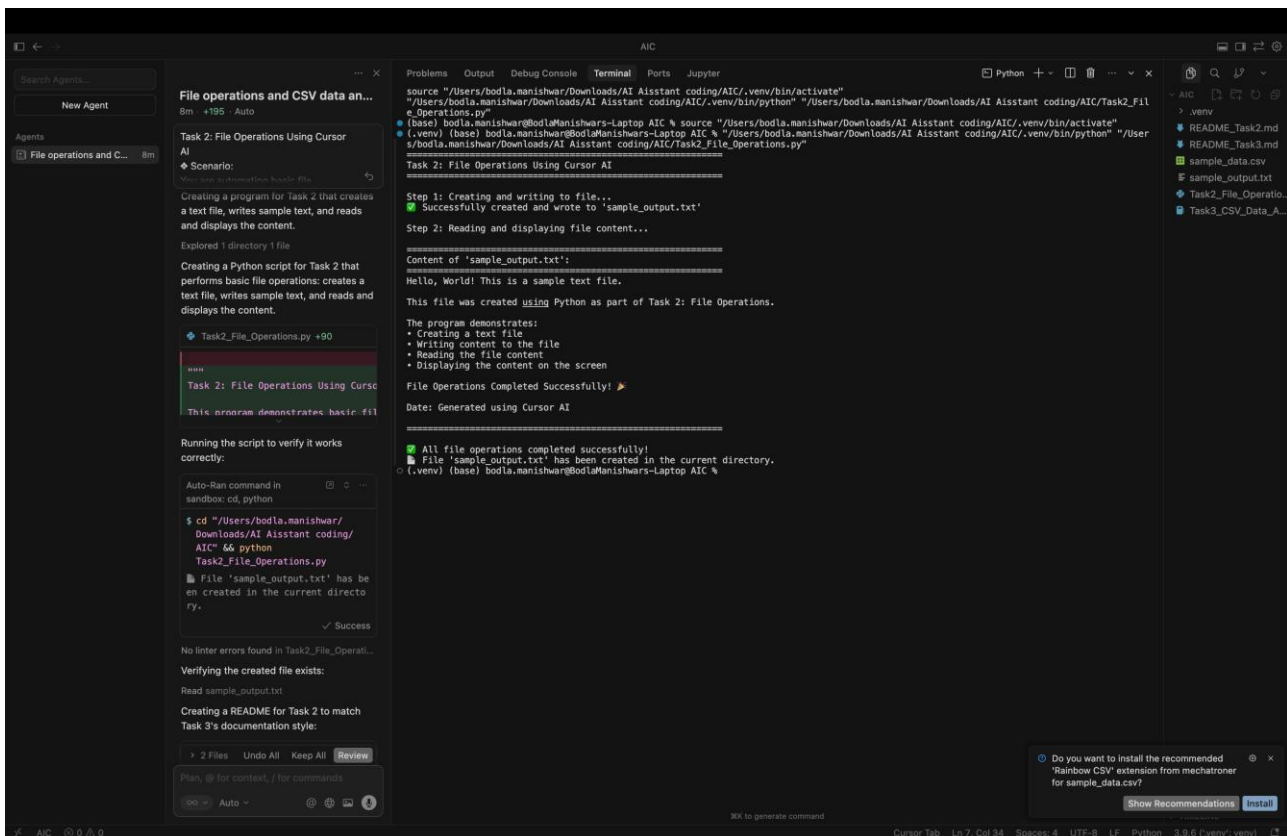
PROMPT:

Generate a simple Python program that demonstrates basic file operations. The program should create a text file, write some sample text into it, then read the content from the file and display it on the screen.

CODE:



OUTPUT:



CODE EXPLANATION:

This Python program demonstrates basic file operations by creating a text file, writing sample content to it, and then reading and displaying that content on the screen. It uses separate functions for writing and reading files to keep the code organized and clear. The program also includes exception handling to manage errors such as file access issues, ensuring smooth execution. The main() function controls the overall flow, and the program runs only when executed directly, making it a simple and effective example of file handling in Python.

Q)Task 3: CSV Data Analysis

❖ Scenario:

You are processing structured data from a CSV file.

❖ Task:

Use Gemini in Colab to read a CSV file and calculate mean, min, and max.

❖ Expected Output:

➤ Correct output

➤ Screenshot

PROMPT:

Write Python code in Google Colab to read a CSV file and calculate mean, minimum, and maximum values using pandas.

CODE:

Search Agents...

New Agent

Agents

CSV file statistical an... 13m

Task 3: CSV Data Analysis

Scenario: You are processing structured data from a CSV file.

Task: Use Gemini in Colab to read a CSV file and calculate mean, min, and max.

Expected Output: Correct output

Screenshot

3. Replace YOUR_API_KEY_HERE in the configuration cell

4. Run all cells — the notebook will calculate and display mean, min, and max values

The notebook is ready to use. Run it in Colab and take a screenshot of the final output showing the statistical analysis results.

3 Files

Task3_CSV_Data_An... +14 -0

sample_data.csv +9 -1

README_Task3.md +93 -1

Reject, suggest, follow up?

Task3_CSV_Data_Analysis.ipynb

Task 3: CSV Data Analysis with Gemini > Step 7: Final Output Summary

+ Code + Markdown | Run All Restart Clear All Outputs | Jupyter Variables Outlinevenv (Python 3.9.6)

Step 1: Install Required Libraries

Install required packages
%pip install -q google-generativeai pandas numpy

Step 2: Import Libraries and Setup

import pandas as pd
import numpy as np
import google.generativeai as genai
import os
from IPython.display import display, HTML

print("Libraries imported successfully!")

Libraries imported successfully!

Do you want to install the recommended 'Rainbow CSV' extension from mechatroner for sample_data.csv?

Show Recommendations Install

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CSV file statistical an... 14m

Task 3: CSV Data Analysis

Scenario: You are processing structured data from a CSV file.

Task: Use Gemini in Colab to read a CSV file and calculate mean, min, and max.

Expected Output: Correct output

Screenshot

showing all results

Ready for Colab — can be uploaded and run directly

Expected Output:

The notebook produces a final summary table like:

Column	Mean	Min	Max
Age	32.75	25	45
Salary	63750.00	50000	80000
Score	89.63	85	95

To Use:

1. Upload the notebook to Google Colab

2. Get your Gemini API key from Google AI Studio

3. Replace YOUR_API_KEY_HERE in the configuration cell

4. Run all cells — the notebook will calculate and display mean, min, and max values

The notebook is ready to use. Run it in Colab and take a screenshot of the final output showing the statistical analysis results.

3 Files

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+ Code + Markdown | Run All Restart Clear All Outputs | Jupyter Variables Outlinevenv (Python 3.9.6)

Step 3: Configure Gemini API

Note: You need to get your Gemini API key from Google AI Studio

Configure Gemini API
Option 1: Set your API key here (replace with your actual key)
GENEIMI_API_KEY = "YOUR_API_KEY_HERE"

Option 2: Or use environment variable
GENEIMI_API_KEY = os.getenv("GENEIMI_API_KEY")

Configure the API
genai.configure(api_key=GENEIMI_API_KEY)

print("Gemini API configured successfully!")

Gemini API configured successfully!

Step 4: Upload CSV File

Upload your CSV file using the file uploader below, or use a sample CSV file.

Read the CSV file
csv_file = "sample_data.csv" # Change this to your uploaded file name

If you uploaded a file, uncomment and use:
csv_file = list(uploaded.keys())[0]

df = pd.read_csv(csv_file)

print("CSV file loaded successfully!")
print(f"Shape: {df.shape}")
print(f"First few rows:")
display(df.head())

CSV file loaded successfully!

Shape: (8, 4)

First few rows:

	Name	Age	Salary	Score
0	Alice	25	50000	85
1	Bob	30	60000	90
2	Charlie	35	70000	88
3	Diana	28	55000	92
4	Eve	32	65000	87

Step 5: Traditional Statistical Analysis (Baseline)

Do you want to install the recommended 'Rainbow CSV' extension from mechatroner for sample_data.csv?

Show Recommendations Install

Search Agents...

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Agents

CSV file statistical an... 14m

Task 3: CSV Data Analysis

15m +624 Auto

Task 3: CSV Data Analysis

Scenario:
You are processing structured data from a CSV file.

Task:
Use Gemini in Colab to read a CSV file and calculate mean, min, and max.

Expected Output:
> Correct output
> Screenshot

Salary 63750.00 50000 80000
Score 89.63 85 95

To Use:
1. Upload the notebook to Google Colab
2. Get your Gemini API key from Google AI Studio
3. Replace YOUR_API_KEY_HERE in the configuration cell
4. Run all cells — the notebook will calculate and display mean, min, and max values
The notebook is ready to use. Run it in Colab and take a screenshot of the final output showing the statistical analysis results.

Task3_CSV_Data_An... +14 -6
sample_data.csv +9 -1
README_Task3.md +93 -1

Reject, suggest, follow up?

Task3_CSV_Data_Analysis.ipynb

Task3_CSV_Data_Analysis.ipynb > M4 Task 3: CSV Data Analysis with Gemini > M4 Step 7: Final Output Summary

+ Code + Markdown ▶ Run All ⏮ Restart ⏭ Clear All Outputs Jupyter Variables Outline ...

venv (Python 3.9.6)

Step 5: Traditional Statistical Analysis (BaseLine)
First, let's calculate mean, min, and max using traditional methods for comparison.

Calculate statistics for numeric columns only
numeric_cols = df.select_dtypes(include=[np.number]).columns

print("=" * 60)
print("TRADITIONAL STATISTICAL ANALYSIS")
print("=" * 60)

stats_df = pd.DataFrame({
 'Column': numeric_cols,
 'Mean': [df[col].mean() for col in numeric_cols],
 'Min': [df[col].min() for col in numeric_cols],
 'Max': [df[col].max() for col in numeric_cols]
})

display(stats_df)

print("\nDetailed Statistics:")
print(df[numeric_cols].describe())

0.0s Python

TRADITIONAL STATISTICAL ANALYSIS

	Column	Mean	Min	Max
0	Age	32.750	25	45
1	Salary	63750.000	50000	80000
2	Score	89.625	85	95

Detailed Statistics:

	Age	Salary	Score
count	8.000000	8.000000	8.000000
mean	32.750000	63750.000000	89.625000
std	6.408099	9895.886591	3.113009
min	25.000000	50000.000000	85.000000
25%	28.750000	57250.000000	87.750000
50%	31.000000	62500.000000	89.500000
75%	35.750000	70500.000000	91.250000
max	45.000000	80000.000000	95.000000

Do you want to install the recommended 'Rainbow CSV' extension from mechatroner for sample_data.csv?
Show Recommendations Install

Search Agents...

New Agent

Agents

CSV file statistical an... 16m

Task 3: CSV Data Analysis

16m +624 Auto

Task 3: CSV Data Analysis

Scenario:
You are processing structured data from a CSV file.

Task:
Use Gemini in Colab to read a CSV file and calculate mean, min, and max.

Expected Output:
> Correct output
> Screenshot

Salary 63750.00 50000 80000
Score 89.63 85 95

To Use:
1. Upload the notebook to Google Colab
2. Get your Gemini API key from Google AI Studio
3. Replace YOUR_API_KEY_HERE in the configuration cell
4. Run all cells — the notebook will calculate and display mean, min, and max values
The notebook is ready to use. Run it in Colab and take a screenshot of the final output showing the statistical analysis results.

Task3_CSV_Data_An... +14 -6
sample_data.csv +9 -1
README_Task3.md +93 -1

Reject, suggest, follow up?

Task3_CSV_Data_Analysis.ipynb

Task3_CSV_Data_Analysis.ipynb > M4 Task 3: CSV Data Analysis with Gemini > M4 Step 7: Final Output Summary

+ Code + Markdown ▶ Run All ⏮ Restart ⏭ Clear All Outputs Jupyter Variables Outline ...

venv (Python 3.9.6)

Step 6: Gemini-Powered Analysis
Now, let's use Gemini to analyze the CSV data and calculate statistics.

Prepare data for Gemini
Convert DataFrame to string format
data_preview = df.head(10).to_string()
data_summary = f"\nData shape: {df.shape}\n"
data_summary += f"\nColumns: {list[Any](df.columns)}\n"
data_summary += f"\nNumeric columns: {list[Any](numeric_cols)}\n"

print("Data prepared for Gemini analysis")

0.0s Python

Data prepared for Gemini analysis

Step 7: Final Output Summary
Mean, Min, Max Values:

Final comprehensive summary
print("=" * 70)
print("FINAL STATISTICAL ANALYSIS - MEAN, MIN, MAX")
print("=" * 70)

final_stats = pd.DataFrame({
 'Column': numeric_cols,
 'Mean': [round(df[col].mean(), 2) for col in numeric_cols],
 'Min': [df[col].min() for col in numeric_cols],
 'Max': [df[col].max() for col in numeric_cols]
})

Display with better formatting
display(HTML(final_stats.to_html(index=False, classes='table table-striped')))

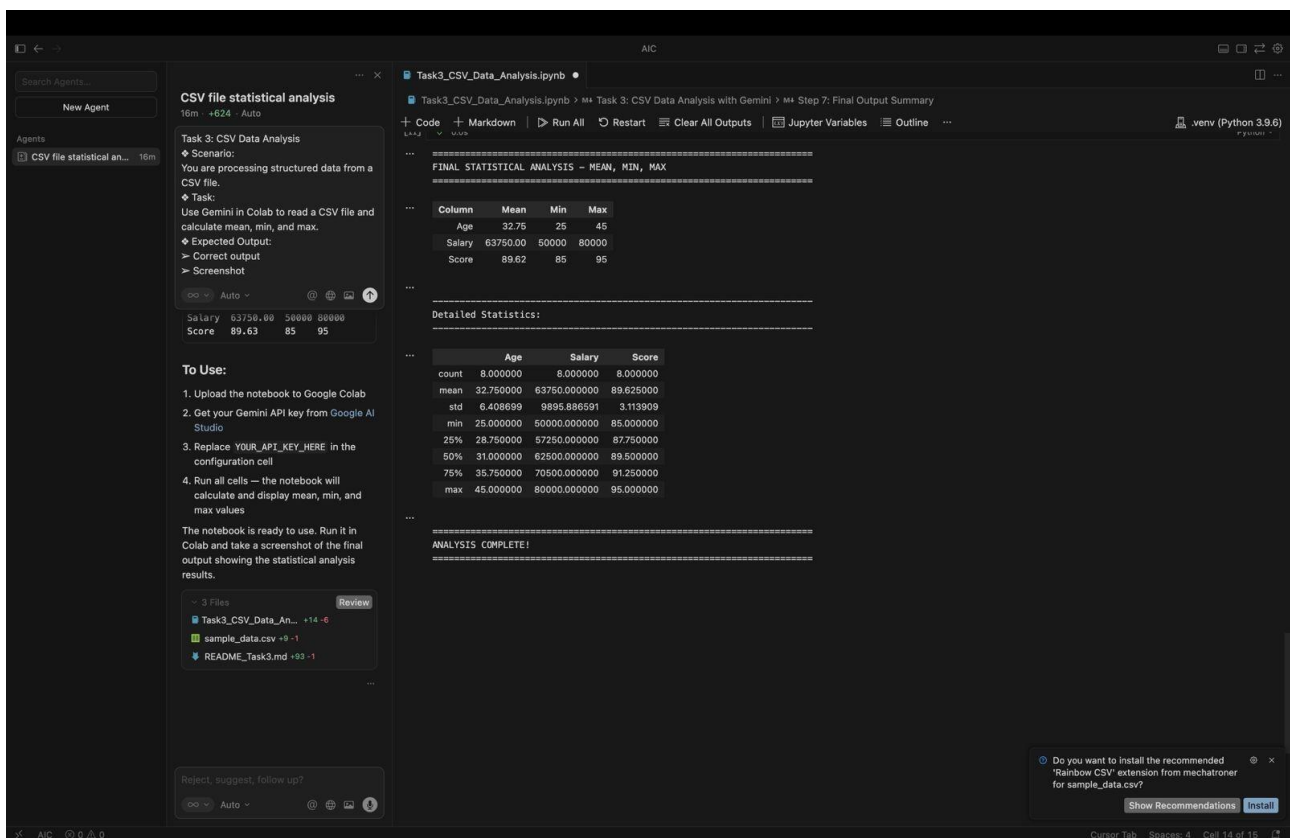
print("\n" + "=" * 70)
print("Detailed Statistics:")
print("=" * 70)
display(df[numeric_cols].describe())

print("\n" + "=" * 70)
print("ANALYSIS COMPLETE!")
print("\n" * 70)

0.0s Python

Do you want to install the recommended 'Rainbow CSV' extension from mechatroner for sample_data.csv?
Show Recommendations Install

OUTPUT:



CODE EXPLANATION:

This code performs statistical analysis on numeric columns of a DataFrame (df). First, it identifies all columns that contain numerical data using `select_dtypes(include=[np.number])`. Then, for each numeric column, it calculates the mean, minimum, and maximum values and stores them in a new DataFrame called `stats_df`. This DataFrame is displayed to show a clean summary of basic statistics.

Q)Task 4: Sorting Lists – Manual vs Built-in

❖ Scenario:

You are reviewing algorithm choices for efficiency.

❖ Task:

Use Gemini to generate:

- Bubble sort
- Python's built-in `sort()`
- Compare both implementations.

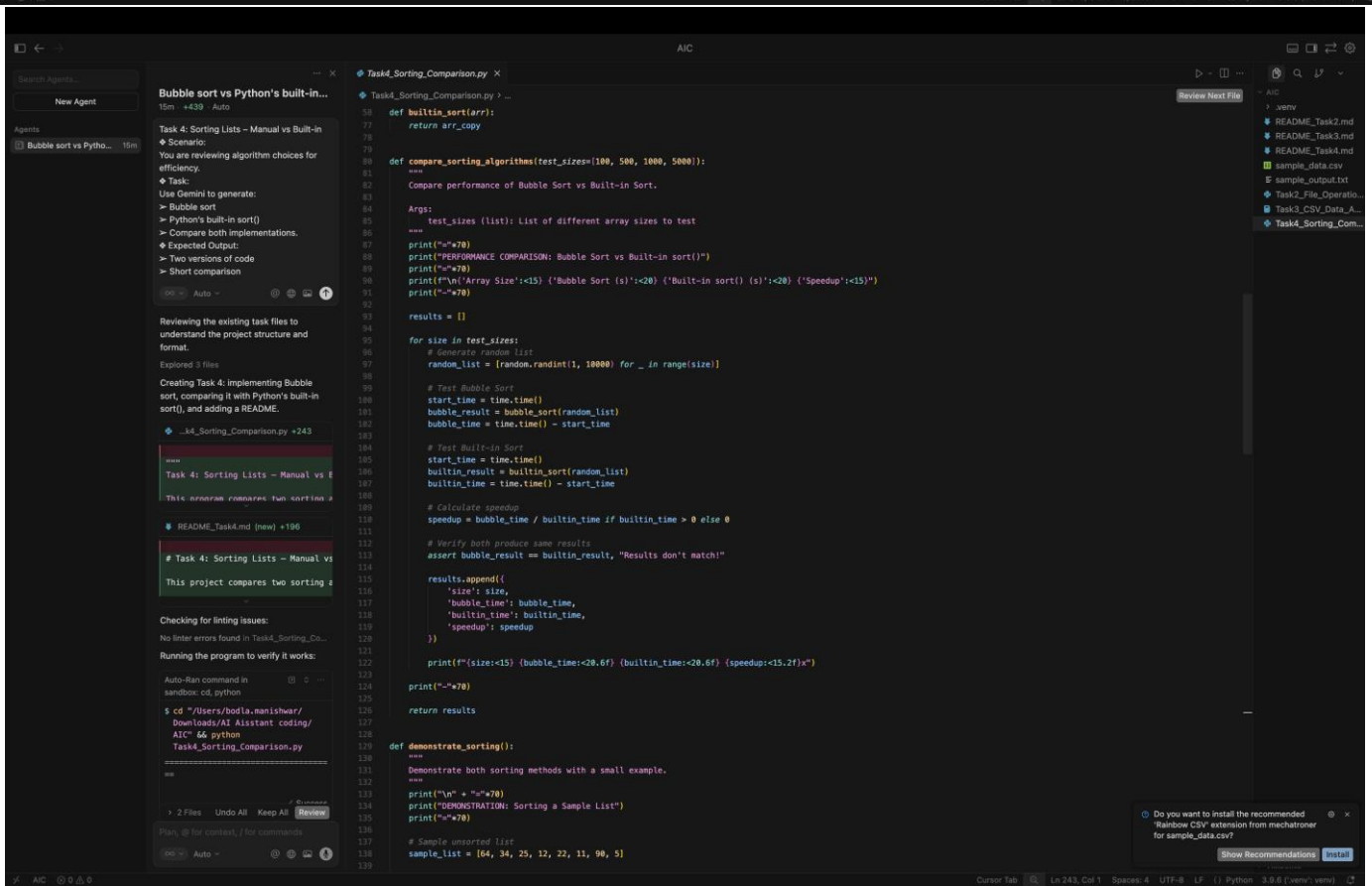
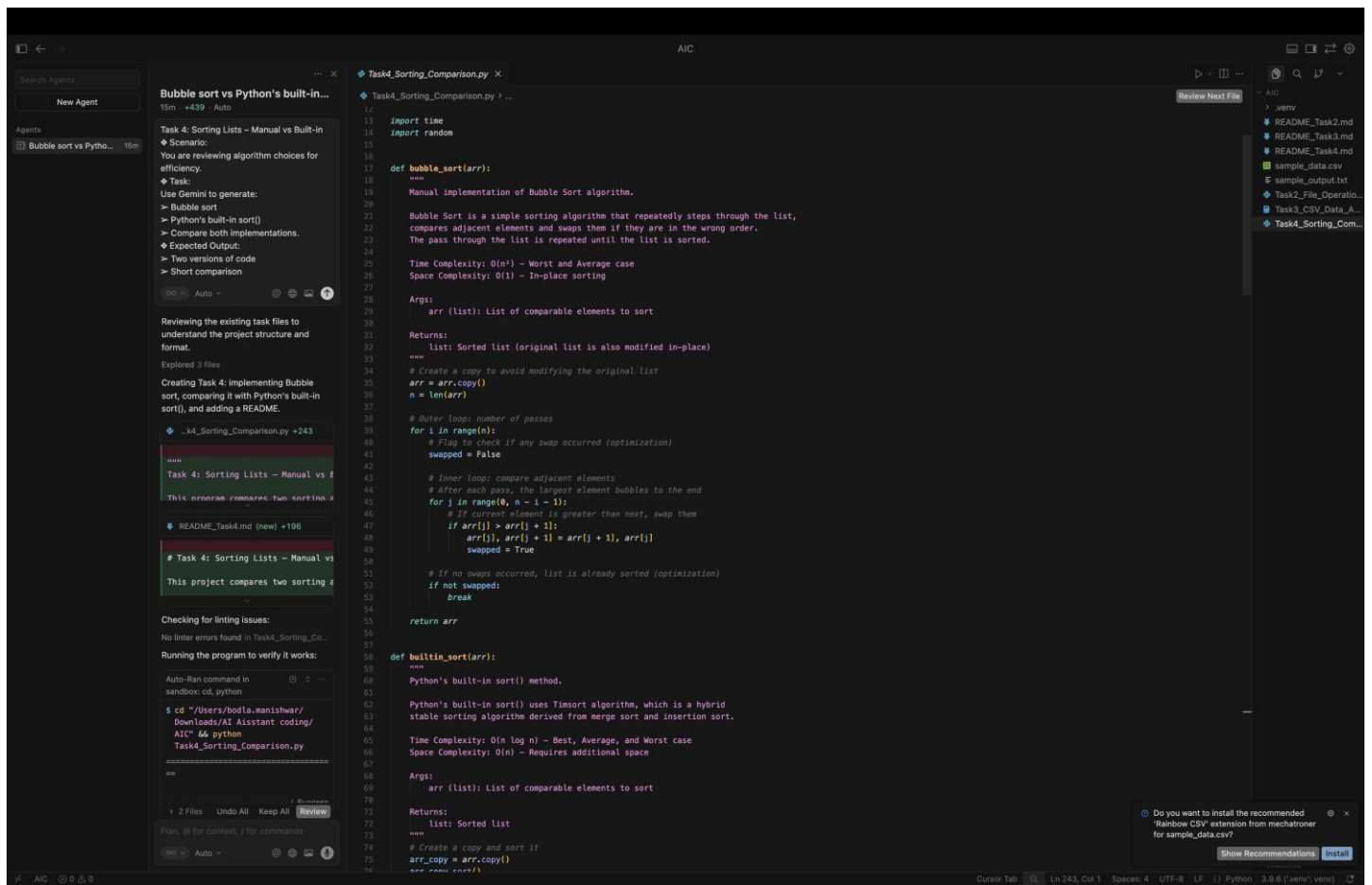
❖ Expected Output:

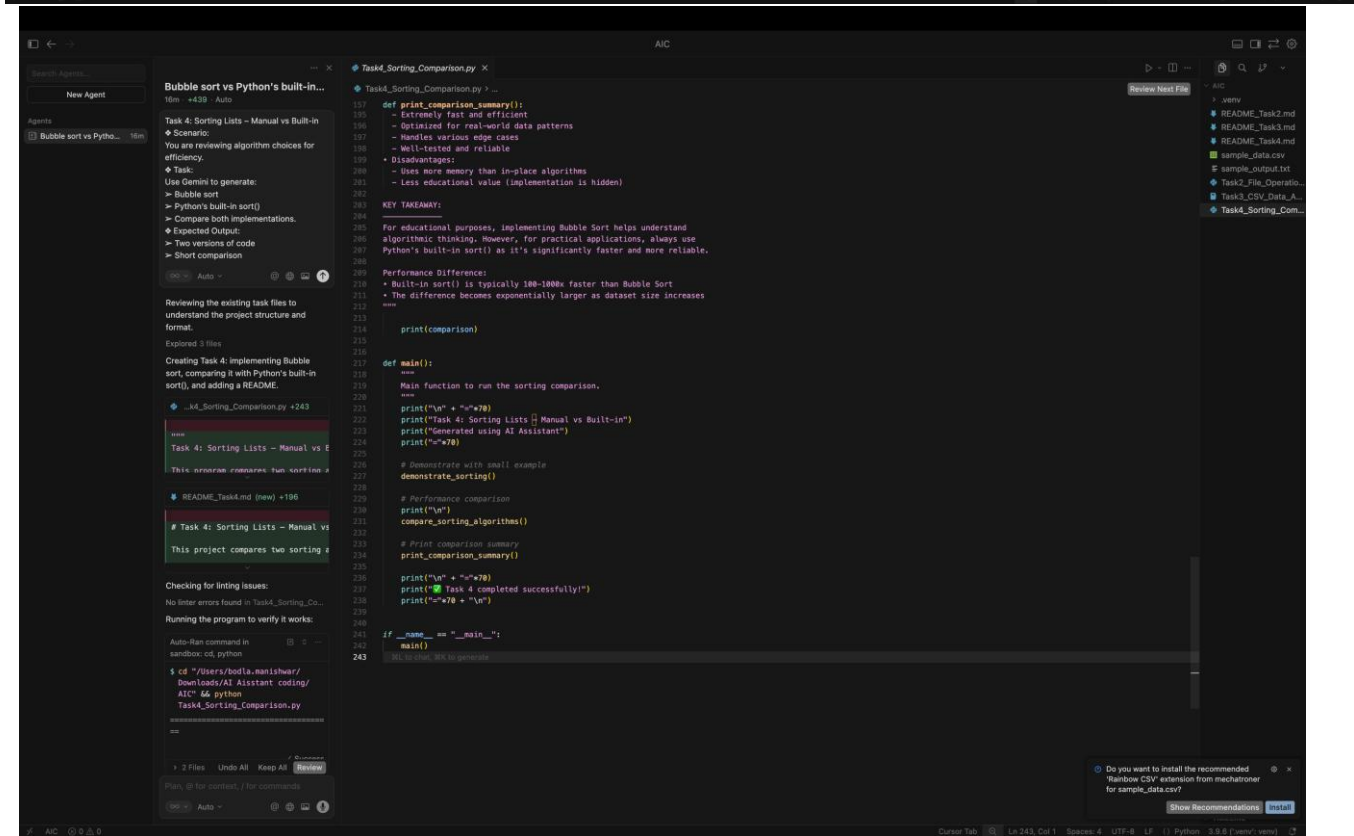
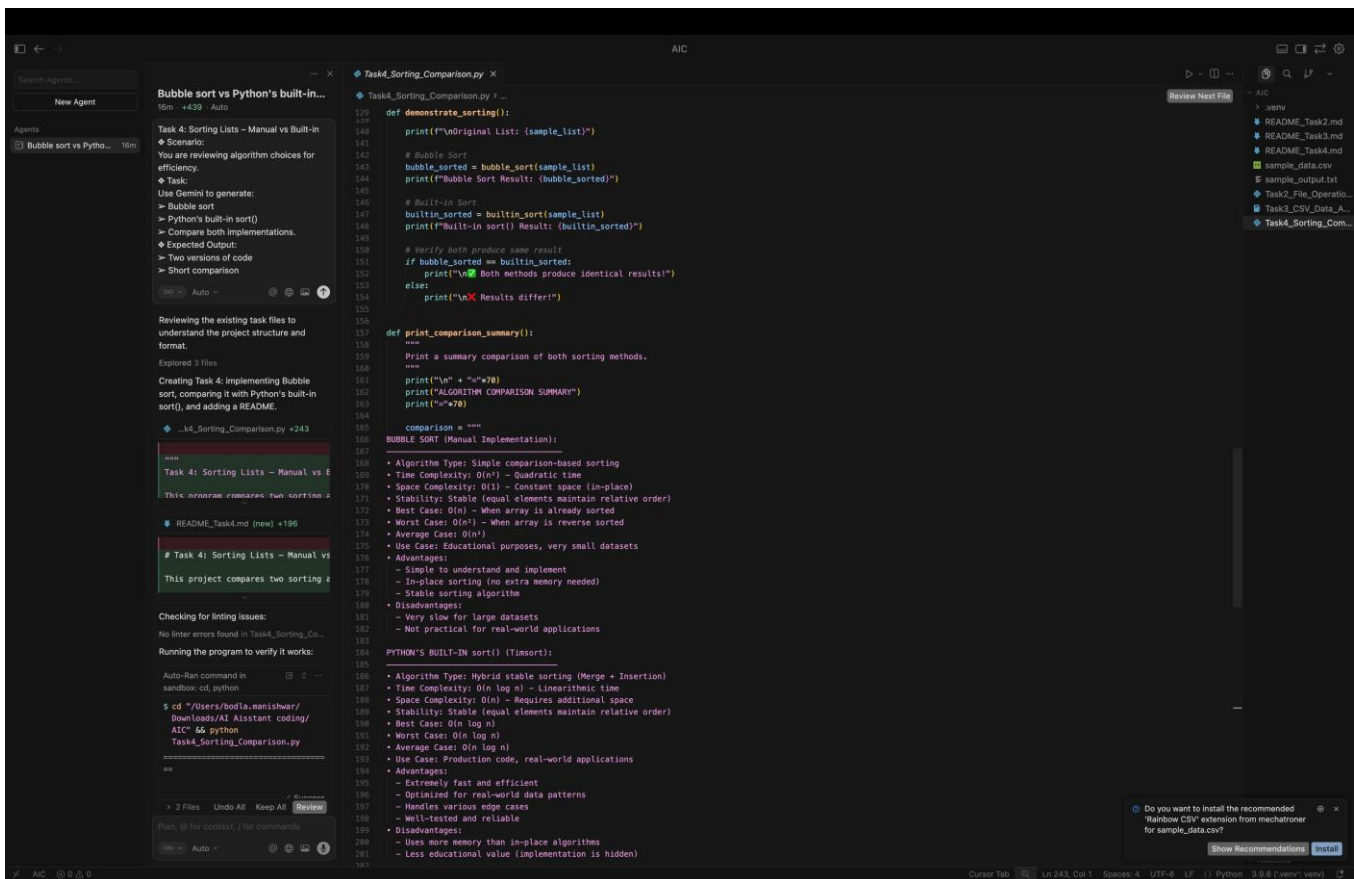
- Two versions of code
- Short comparison

PROMPT:

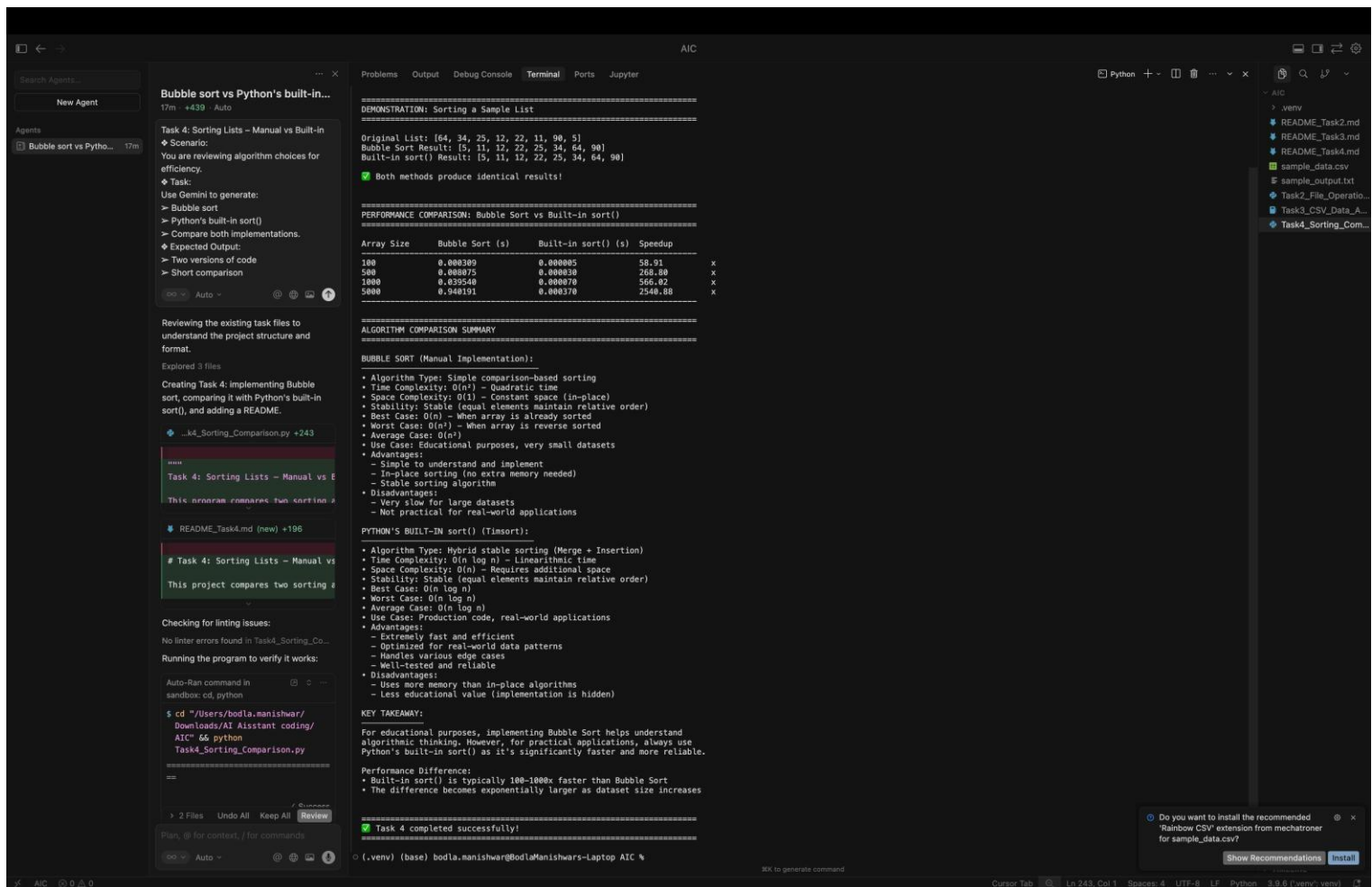
Generate Python code to sort a list using bubble sort and Python's built-in `sort()` method. Show both implementations clearly and provide a short comparison explaining their efficiency and usage.

CODE:





OUTPUT:



CODE EXPLANATION:

This program compares Bubble Sort and Python's built-in sort(). Bubble Sort manually compares and swaps elements to arrange them in order, but it is slow for large lists because it has $O(n^2)$ time complexity. Python's built-in sort() uses an efficient algorithm and sorts data much faster with $O(n \log n)$ time complexity. The program measures execution time for both methods and shows that the built-in sort is much faster and more suitable for real-world use.