

AI Assisted Coding LAB ASSIGNMENT-5.4

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Batch:14

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Task Description #1:

- Prompt GitHub Copilot to generate a Python script that collects user data (e.g., name, age, email). Then, ask Copilot to add comments on how to anonymize or protect this data.

PROMPT

```
# Generate a Python script that collects user data such as  
name, age, and email.
```

```
# Add inline comments explaining how to protect or  
anonymize this data,
```

```
# such as hashing email addresses, avoiding plain-text  
storage,
```

```
# and following basic privacy best practices.
```

The screenshot shows a GitHub repository named "user-data-collection" with a file "user_data_gathering.py". The code implements various security measures:

- Hashing passwords using SHA-256.
- Validating user input for age and email.
- Using environment variables for sensitive data.
- Minimizing data collection by asking for only what's necessary.
- Ensuring secure user data collection.
- Collecting name and validating age.
- Collecting email and validating it.
- Collecting password and validating its length.
- Creating a user record with minimal sensitive data.

The repository also includes a "README.md" file with instructions and a "LICENSE" file.

The screenshot shows a developer's environment with several windows open:

- Code Editor:** The main window displays a Python script titled "use_data_privacy.py". The code implements a class "UserManager" with methods for adding users and displaying user data. It includes annotations for security principles like "PLAIN-TEXT" and "ENCRYPTED".
- Terminal:** A terminal window at the bottom shows the command "python use_data_privacy.py" being run.
- Output:** A "OUTPUT" tab shows the execution results, including error messages about file permissions and imports.
- File Explorer:** Shows the project structure with files like "use_data_privacy.py", "use_data_privacy.pyc", and "use_data_privacy.pycn".
- Search:** A search bar at the top right is set to "CHAT".
- Help:** A help menu is visible at the top.

Expected Output #1:

- A script with inline Copilot-suggested code and comments explaining how to safeguard or anonymize user information (e.g., hashing emails, not storing data unencrypted).

```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS + - | X powers... Python

Select option (1-4): & C:/Users/Sreeshma/AppData/Local/Programs/Python/Python310/python.exe "c:/Users/Sreeshma/Documents/AI ASS/user_data_privacy.py"
ERROR: Invalid option. Please select 1-4.

--- OPTIONS ---
1. Add new user
2. View anonymized data
3. Save data securely
4. Exit

Select option (1-4): 1

== Secure User Data Collection ==
Enter your name (will be stored): Nitish
Enter your age: 20
Enter your email (will be hashed for privacy): nitishrajkond@gmail.com
Enter a password (hidden for security):
ERROR: Password must be at least 8 characters.

--- OPTIONS ---
1. Add new user
2. View anonymized data
3. Save data securely
4. Exit
```

Task Description #2:

- Ask Copilot to generate a Python function for sentiment analysis.

Then prompt Copilot to identify and handle potential biases in the data.

PROMPT: # Generate a Python function for sentiment analysis.

Add comments or code to identify and reduce potential biases in the data,

such as removing offensive terms, balancing positive and negative samples,

and avoiding biased language in predictions.

Expected Output #2:

- Copilot-generated code with additions or comments addressing bias mitigation strategies (e.g., balancing dataset, removing offensive terms).

Task Description #3:

- Use Copilot to write a Python program that recommends products based on user history. Ask it to follow ethical guidelines like transparency and fairness

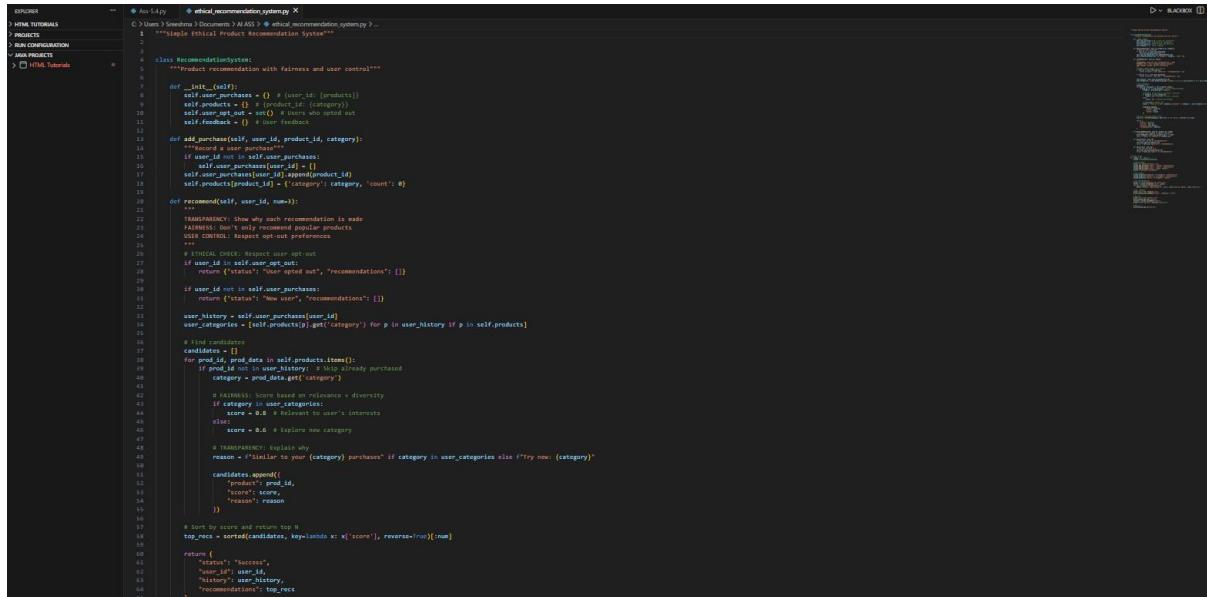
PROMPT: # Generate a Python program that recommends products based on user purchase history.

Follow ethical AI guidelines such as transparency, fairness, and user control.

Add comments explaining how recommendations are generated,

avoid favoritism toward only popular products,

and allow users to give feedback or opt out of recommendations.



The screenshot shows a code editor with a Python file named `ethical_recommendation_system.py`. The code implements a recommendation system with ethical considerations. It includes functions for adding purchases, recommending products, and calculating scores based on popularity, relevance, and diversity. The code is annotated with comments explaining its logic and ethical principles like transparency and fairness.

```
#!/usr/bin/env python3
# ethical_recommendation_system.py

class RecommendationSystem:
    """Implements ethical product recommendation System"""

    def __init__(self):
        self.user_purchases = {} # (user_id: [products])
        self.products = {} # product_id: (category)
        self.user_categories = {} # user_id: [categories]
        self.feedback = {} # user_feedback

    def add_purchase(self, user_id, product_id, category):
        """Record a user purchase"""
        if user_id not in self.user_purchases:
            self.user_purchases[user_id] = []
        self.user_purchases[user_id].append(product_id)
        self.user_categories[user_id].append(category)
        self.products[product_id] = {'category': category, 'count': 0}

    def recommend(self, user_id, num=10):
        """TRANSPARENCY: show why each recommendation is made
        Fairness: Don't only recommend popular products
        User control: Respect user's opt-out
        """
        if user_id not in self.user_purchases:
            return {"status": "User opted out", "recommendations": []}
        if user_id not in self.user_categories:
            return {"status": "New user", "recommendations": []}

        user_history = self.user_purchases[user_id]
        user_categories = self.products.get('category') for p in user_history if p in self.products

        # Find candidates
        candidates = []
        for prod_id, prod_data in self.products.items():
            if prod_id not in user_history: # Skip already purchased
                category = prod_data.get('category')
                if category in user_categories: # relevance + diversity
                    if category in user_categories:
                        score = 0.8 # Relevant to user's interests
                    else:
                        score = 0.6 # Explore new category
                else:
                    reason = "Similar to your [category] purchases" if category in user_categories else "[try now] (category)"
                    candidates.append({
                        "product": prod_id,
                        "score": score,
                        "reason": reason
                    })
        top_rec = sorted(candidates, key=lambda x: x['score'], reverse=True)[num]

        return {
            "user_id": user_id,
            "user_ip": user_ip,
            "history": user_history,
            "recommendations": top_rec
        }
```

```

# Example usage
if __name__ == "__main__":
    system = RecommendationSystem()

    # Add purchases
    print("--- Adding Purchases ---")
    system.add_purchase("user1", "laptop", "electronics")
    system.add_purchase("user1", "monitor", "electronics")
    system.add_purchase("user2", "book", "books")
    print("Purchases recorded")

    # Add products
    system.products["keyboard"] = {"category": "Electronics"}
    system.products["monitor"] = {"category": "Electronics"}
    system.products["novel1"] = {"category": "Books"}

    # Get recommendations
    print("--- Recommendations for user1 ---")
    result1 = system.recommend("user1", num=2)
    for rec in result1["recommendations"]:
        print(f"Product: {rec['product']}, Score: {rec['score']}, Reason: {rec['reason']}")

    # Give feedback
    print("--- User Feedback ---")
    print(system.give_feedback("user1", "keyboard", True))

    # Opt out
    print("--- User Control ---")
    print(system.opt_out("user1"))
    result2 = system.recommend("user1")
    print("After opt-out: ", result2["status"])

    # Opt in
    print(system.opt_in("user1"))

```

Expected Output #3:

- Copilot suggestions that include explanations, fairness checks (e.g., avoiding favoritism), and user feedback options in the code.

```

--- Adding Purchases ---
Purchases recorded
PS C:\Users\greenba\Downloads\HTML Tutorials> & C:/Users/greenba/AppData/Local/Programs/Python/Python310/python.exe "C:/Users/greenba/Documents/AI AI/ethical_recommendation_system.py"
PS C:\Users\greenba\Downloads\HTML Tutorials> & C:/Users/greenba/AppData/Local/Programs/Python/Python310/python.exe "C:/Users/greenba/Documents/AI AI/ethical_recommendation_system.py"

--- Adding Purchases ---
Purchases recorded

--- Recommendations for user1 ---
Product: laptop, Score: 8.8, Reason: Stellar to your Electronics purchases
Product: monitor, Score: 8.8, Reason: Stellar to your Electronics purchases

--- User Feedback ---
Thanks for feedback on keyboard

--- User Control ---
user1 opted out of recommendations
After opt-out: user1 opted out
user1 opted out of recommendations
PS C:\Users\greenba\Downloads\HTML Tutorials>

```

Task Description #4:

- Prompt Copilot to generate logging functionality in a Python web application. Then, ask it to ensure the logs do not record sensitive information.

PROMPT: # Generate logging functionality for a Python web application.

Ensure logs do NOT store sensitive information such as passwords,

emails, or personal identifiers.

Add comments explaining ethical logging practices and privacy protection.

```
BUKU -- ethical_logging.py ethical recommendation systempy ethical_loggingpy X
C:\Users>Somedemo>Documents>AI\ADS>ethical_loggingpy >...
1 #!/usr/bin/python
2 # Example Ethical Logging for web Applications
3
4 import logging
5 import re
6
7
8 class PrivacyFilter(logging.Filter):
9     """Remove sensitive data from logs"""
10    def filter(self, record):
11        """Mask password, emails, tokens, cards before logging"""
12        msg = self.format(record)
13
14        # PRIVACY: Mask password
15        msg = re.sub(r'password=[^"]*', 'password=""REDACTED""', msg, flags=re.IGNORECASE)
16
17        # PRIVACY: Mask emails (show domain only)
18        msg = re.sub(r'([^\@]+\@[^\@]+\.[^\@]+)', '[REDACTED]', msg)
19
20        # PRIVACY: Mask API keys and tokens
21        msg = re.sub(r'([^\@]+\:[^\@]+\@[^\@]+\:[^\@]+\:[^\@]+)', ':[REDACTED]:[REDACTED]:[REDACTED]:[REDACTED]', msg, flags=re.IGNORECASE)
22
23        # PRIVACY: Mask credit card numbers (show last 4 digits)
24        msg = re.sub(r'([^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+)', '([CCN]\w{4})', msg)
25
26        # PRIVACY: Mask phone numbers (show last 4 digits)
27        msg = re.sub(r'([^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+\:[^\@]+)', '([PNEUT]\w{4})', msg)
28
29
30        record.msg = msg
31
32    return True
33
34
35 def setup_logger(name, log_file='app.log'):
36     """Setup logger with file and console protection"""
37     logger = logging.getLogger(name)
38     logger.setLevel(logging.INFO)
39
40     # Add privacy filter
41     privacy_filter = PrivacyFilter()
42
43     # Console Handler
44     console_handler = logging.StreamHandler()
45     console_handler.addFilter(privacy_filter)
46     formatter = logging.Formatter('%(asctime)s - %(name)s - %(levelname)s - %(message)s')
47     console_handler.setFormatter(formatter)
48     logger.addHandler(console_handler)
49
50     # File Handler
51     if log_file:
52         file_handler = logging.FileHandler(log_file)
53         file_handler.addFilter(privacy_filter)
54         file_handler.setLevel(logging.DEBUG)
55         logger.addHandler(file_handler)
56
57     # PRIVACY: Restrict file permissions (owner read/write only)
58     import os
59     os.chmod(log_file, 0600)
60
61     return logger
62
63
64 def log_user_action(logger, action, user_id, **safe_details):
65     """Log user action with only safe fields"""
66     msg = f'ACTION: {action} | user: {user_id}'
67     if safe_details:
68        msg += f' | {safe_details}'
69        logger.info(msg)
70
71    # Example usage
72    if __name__ == "__main__":
73        print("---- simple Ethical logging demo ----")
74
75        logger = setup_logger('app', log_file='app.log')
76
77        print("Test 1: Password Masking")
78        logger.info("login with password=SecurePass123")
79
80        print("Test 2: Email Masking")
81        logger.info("Send email to user@example.com")
82
```

```
BUKU -- ethical_logging.py ethical recommendation systempy ethical_loggingpy X
C:\Users>Somedemo>Documents>AI\ADS>ethical_loggingpy >...
12 def setup_logger(name, log_file='app.log'):
13
14    return logger
15
16
17 def log_user_action(logger, action, user_id, **safe_details):
18    """Log user action with only safe fields"""
19    msg = f'ACTION: {action} | user: {user_id}'
20    if safe_details:
21        msg += f' | {safe_details}'
22        logger.info(msg)
23
24
25    # Example usage
26    if __name__ == "__main__":
27        print("---- simple Ethical logging demo ----")
28
29        logger = setup_logger('app', log_file='app.log')
30
31        print("Test 1: Password Masking")
32        logger.info("login with password=SecurePass123")
33
34        print("Test 2: Email Masking")
35        logger.info("Send email to user@example.com")
36
37        print("Test 3: API Key Masking")
38        logger.info("API key: sk_1live_1234abcde")
39
40        print("Test 4: Credit Card Masking")
41        logger.info("Payment with card 4321-1234-5678-9999")
42
43        print("Test 5: User Action Logging")
44        log_user_action(logger, "purchase", "user_123", status="success", amount=99.99)
45
46        print("User ID: " + str(user_id))
47        print("---- LOGGING PRACTICES ----")
48        print("1. NO FILTER: Mask passwords, emails, tokens, cards")
49        print("2. MINIMAL DATA: Only log necessary information")
50        print("3. SECURE FILES: Set permissions to 600 (owner only)")
51        print("4. DEBUG LEVEL LOG FOR DEVELOPMENT ONLY")
52        print("5. NO SENSITIVE DATA IN LOGS")
53
54
```

Expected Output #4:

- Logging code that avoids saving personal identifiers (e.g., passwords, emails), and includes comments about ethical logging practices.

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PAGES
Test 5: User Action Logging
2024-01-29 18:20:55,566 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}

ETHICAL LOGGING PRACTICES:
1. PRIVACY FILTER: Mask passwords, emails, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 600 (owner only)
4. AUDIT ACTIONS: Log for auditing and debugging
5. NO SECRETS: Never store sensitive data in logs
2024-01-29 18:20:55,566 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}

ETHICAL LOGGING PRACTICES:
1. PRIVACY FILTER: Mask passwords, emails, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 600 (owner only)
4. AUDIT ACTIONS: Log for auditing and debugging
5. NO SECRETS: Never store sensitive data in logs
2024-01-29 18:20:55,566 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}
5. NO SECRETS: Never store sensitive data in logs
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```

Task Description #5:

- Ask Copilot to generate a machine learning model. Then, prompt it to add documentation on how to use the model responsibly (e.g., explainability, accuracy limits).

PROMPT: Generate a Python machine learning model (including data loading, training, and prediction steps).

Add inline documentation or a README-style comment section explaining how to use the model responsibly, including accuracy limitations, explainability considerations, fairness concerns, and appropriate use cases and restrictions.

```
EXPLORER
> HTML TUTORIALS
> PROJECTS
> RECOMMENDATION
> JAVA PROJECTS
> HTML Tutorials

-- Ass-5-4.py 伦理推荐系统.py ethical_logging.py responsible_ml_model.py

C:\Users\Seethima\Documents\AI\Ass-5> python responsible_ml_model.py

67     recs, reasons = recommend_products(user_id, user_history, product_catalog)
68     for prod, reason in zip(recs, reasons):
69         print(f"(prod['name']) {category}: (prod['category'])) -> (reason)")
70
71     # User feedback and opt-out
72     print("Would you like to provide Feedback or opt out of recommendations?")
73     feedback = input("Type 'feedback' to provide feedback or 'opt out' to stop recommendations: ")
74
75     if feedback.lower() == 'opt out':
76         print("You have opted out of recommendations. Your preferences will be respected.")
77     else:
78         print(f"Thank you for your Feedback: {feedback}")
79
80     # --- Ethical AI Notes ---
81     # - Transparency: Each recommendation includes an explanation.
82     # - Fairness: The system ensures diversity and avoids recommending only from the most frequent category.
83     # - User Control: Users can provide feedback or opt out at any time.
84     # - Regularly audit recommendation logic for bias and update as needed.
85     # A few more required packages are installed
86
87 import sys
88 import subprocess
89
90 def install_if_missing(package):
91     try:
92         __import__(package)
93     except ImportError:
94         print(f"Installing missing package: {package}")
95         subprocess.check_call([sys.executable, "-m", "pip", "install", package])
96
97 # Install 'textblob' if not present
98 install_if_missing('textblob')
99
100 # Sentiment analysis function with bias awareness and mitigation strategies
101 from textblob import TextBlob
102
103 def analyze_sentiment(text):
104     """
105     Analyze the sentiment of the input text.
106     Returns polarity (-1 to 1) and subjectivity (0 to 1).
107
108     Potential sources of bias in training data:
109     - Imbalanced datasets (e.g., more positive than negative samples)
110     - Presence of offensive, discriminatory, or culturally specific terms
111     - Overrepresentation or underrepresentation of certain topics or groups
112
113     Strategies to mitigate bias:
114     - Balance the dataset across sentiment classes and demographic groups
115     - Remove or flag offensive/discriminatory terms during preprocessing
116     - Use diverse and representative data sources
117     - Document keep a list and test for bias regularly
118     - Involve domain experts in dataset curation
119     ...
120
121     # Example using Textblob for simple sentiment analysis
122     blob = TextBlob(text)
123     polarity = blob.sentiment.polarity
124     subjectivity = blob.sentiment.subjectivity
125     return polarity, subjectivity
126
127
128 # Example usage
129 if __name__ == "__main__":
130     user_text = input("Enter text for sentiment analysis: ")
131     polarity, subjectivity = analyze_sentiment(user_text)
132     print(f"Polarity: {polarity}, Subjectivity: {subjectivity}")
133
134
135 # Note: For production, train your own model on a carefully curated dataset and regularly audit for bias.
136 # The above function uses Textblob, which is trained on general-purpose data and may inherit its biases.
```

Expected Output #5:

- Copilot-generated model code with a README or inline documentation suggesting responsible usage, limitations, and fairness considerations.

File Edit Selection View ... ← →

CP LAB ASS

File Structure

area.py

volume.py

check.py

main.py

PyCharm

Output

Console

Terminal

Help

Code

File Data Collection for Android

Save at: simple_ml_mlmodel

Generating a Python machine learning model including the loading, training, and evaluation steps.

Add this entire documentation or a README file to your project to explain how to load and use the trained model. For more information, see the API reference for accuracy limitations, explainability, transparency, and other important topics and see case studies and restrictions.

Generated Python code for this model:

SimpleMLModel

Created by PyCharm. Last modified by PyCharm.

1 Data Loading - Imports all current libraries and defines constants.

2 Train/Test Split - Shows how separate training and testing datasets are used for the model's training.

3 Model Training - Shows how the model is trained using the training data.

4 Evaluation - Shows how accuracy, precision, recall, confusion matrix, and F1 score are calculated.

5 Explainability - Features importance and partial dependence plots.

6 Deployment - Shows how to use the trained model in production.

7 Conclusion - Test vs real-world performance and discuss next steps.

Documentation Cover:

✓ Accuracy Limitations - Test vs real-world performance and discuss next steps.

✓ Explainability - Features importance and partial dependence plots.

✓ Performance - Model training time and deployment performance metrics.

Documentation Concerns:

✗ Accuracy Limitations - When appropriate, include a discussion of the model's strengths and weaknesses.

✗ Use Cases - When appropriate, include a discussion of the model's intended use cases.

✗ Deployment - Whether weight required, continuous monitoring, and so forth.

README Section includes:

• Comprehensive guide for users

• Installation requirements

• Production deployment requirements

• Configuration options

• Reproducible at dev/test

Save at: model_responsible_acq

File

Edit

Selection

View

...

← →

CP LAB ASS

File Edit Selection View ... ← →

File Structure

area.py

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main.py

PyCharm

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...

← →

CP LAB ASS

File Structure

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Code

File Data Collection for Android

Save at: simple_ml_mlmodel

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