

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.

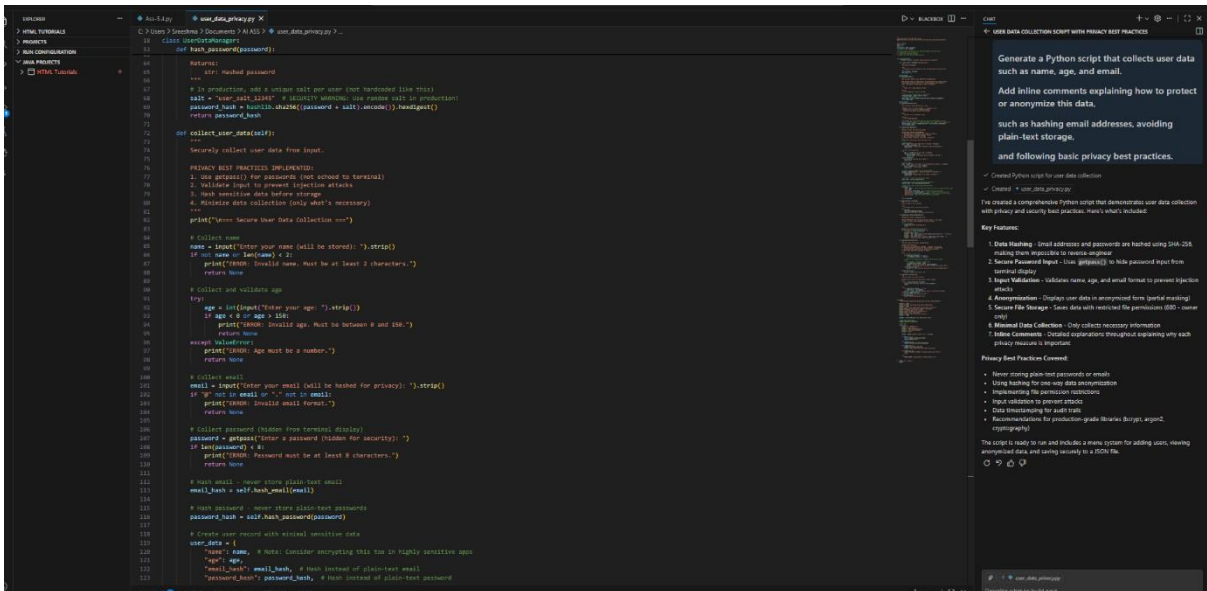
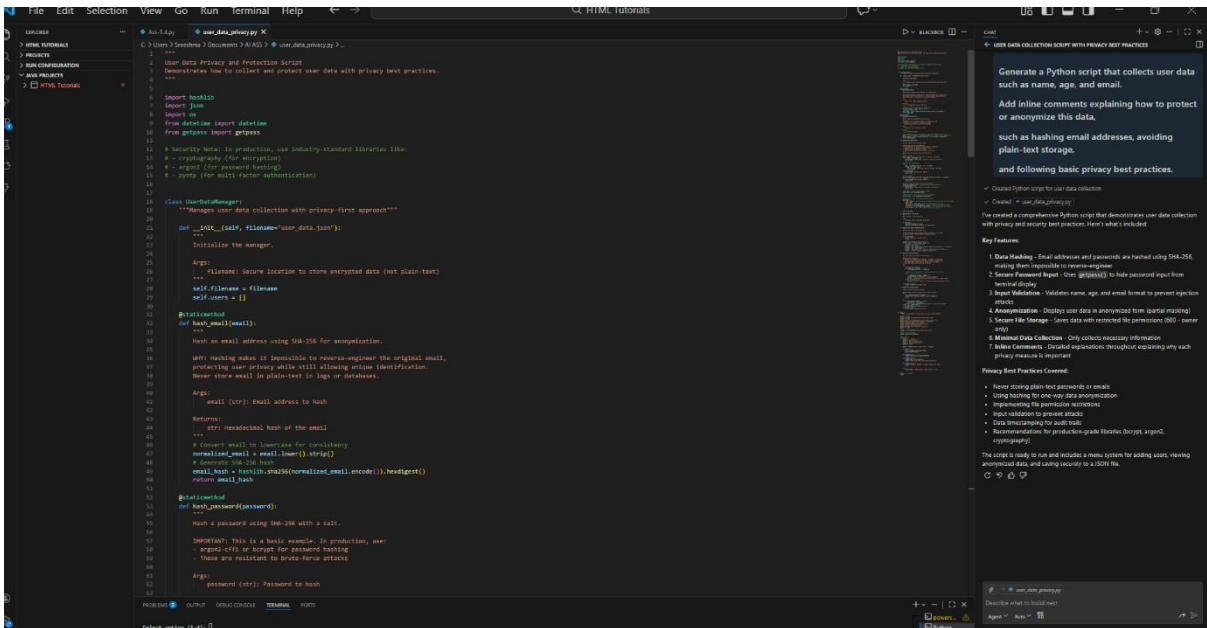
PROMPT

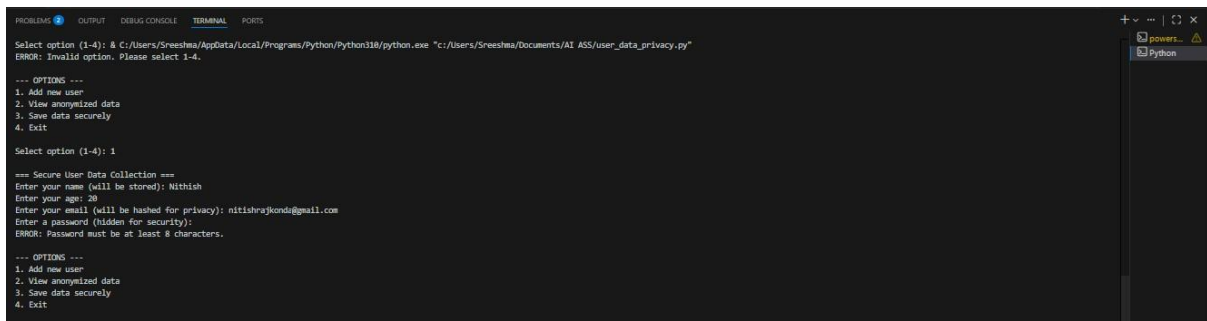
Generate a Python script that collects user data such as name, age, and email.

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.





```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS
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): Nithish
Enter your age: 20
Enter your email (will be hashed for privacy): nithishrajkonda@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.

```
class SentimentClassifier:
    """Sentiment classifier with bias mitigation"""

    def __init__(self):
        """Initialize classifier with bias mitigation"""
        # Define positive and negative words
        self.positive_words = ['great', 'excellent', 'amazing', 'fantastic', 'awesome', 'love', 'best']
        self.negative_words = ['bad', 'terrible', 'awful', 'hate', 'worst', 'dislike']

        # Define words to be removed from the dataset
        self.remove_words = ['bias', 'discrimination', 'offensive']

    def clean_text(self, text):
        """Clean text by removing bias words and stopwords"""
        text = text.lower()
        # Remove stopwords
        for word in self.remove_words:
            text = text.replace(word, '')
        return text

    def analyze(self, text):
        """Analyze sentiment of text"""
        # Calculate sentiment score
        score = self._calculate_score(text)

        # Classify sentiment
        words = self._classify_words(text)

        # Calculate sentiment score
        score = (positive - negative) / (positive + negative)

        # Round score to 1 decimal place
        score = round(score, 1)

        # Classify sentiment
        if score > 0:
            label = 'POSITIVE'
        elif score < 0:
            label = 'NEGATIVE'
        else:
            label = 'NEUTRAL'

        return {'text': text, 'score': score, 'label': label}

    def balance_dataset(self, texts, labels):
        """Balance dataset by removing positive and negative samples"""
        counts = Counter(labels)
        max_count = max(counts.values())

        # Calculate number of samples to remove
        to_remove = {}
        for label in counts:
            to_remove[label] = counts[label] - max_count

        # Remove samples
        balanced_texts, balanced_labels = [], []
        for i, (text, label) in enumerate(zip(texts, labels)):
            if label in to_remove and to_remove[label] > 0:
                balanced_labels.append(label)
                balanced_texts.append(text)
                to_remove[label] -= 1
            else:
                balanced_labels.append(label)
                balanced_texts.append(text)

        print(f'Balanced dataset: {len(balanced_texts)} samples')
        return balanced_texts, balanced_labels

    def train(self):
        """Train the classifier"""
        # Load data
        data = self._load_data()

        # Split data
        train_texts, train_labels = self._split_data(data)

        # Train classifier
        classifier = SentimentClassifier()
        classifier.fit(train_texts, train_labels)

        # Evaluate classifier
        result = classifier.evaluate(train_texts, train_labels)

        # Print results
        print(f'Train Results: {result}')

        # Test classifier
        test_texts, test_labels = self._load_test_data()
        result = classifier.evaluate(test_texts, test_labels)

        # Print results
        print(f'Test Results: {result}')

        # Calculate balanced dataset
        balanced_texts, balanced_labels = self._balance_dataset(train_texts, train_labels)

        # Print results
        print(f'Balanced Dataset Results: {result}')
```

Expected Output #2:

- Copilot-generated code with additions or comments addressing

bias mitigation strategies (e.g., balancing dataset, removing offensive terms).

```
Text: It's okay, nothing special.
Result: 0

PS C:\Users\Sreeshma\Downloads\HTML Tutorials> & C:\Users\Sreeshma\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Sreeshma\Documents\AI ASS\sentiment_analysis_bias.py"
=== Sentiment Analysis ===
Text: This product is amazing and excellent!
Result: {'text': 'This product is amazing and excellent!', 'score': 1.0, 'label': 'POSITIVE'}

Text: I hate this, it's terrible.
Result: {'text': 'I hate this, it's terrible.', 'score': -1.0, 'label': 'NEGATIVE'}

PS C:\Users\Sreeshma\Downloads\HTML Tutorials> & C:\Users\Sreeshma\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Sreeshma\Documents\AI ASS\sentiment_analysis_bias.py"
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=== Sentiment Analysis ===
Text: This product is amazing and excellent!
Result: {'text': 'This product is amazing and excellent!', 'score': 1.0, 'label': 'POSITIVE'}

Text: I hate this, it's terrible.
Text: This product is amazing and excellent!
Result: {'text': 'This product is amazing and excellent!', 'score': 1.0, 'label': 'POSITIVE'}

Text: I hate this, it's terrible.
```

```
=== Dataset Balancing ===
Before: {'POSITIVE': 8, 'NEGATIVE': 2}
After: {'POSITIVE': 2, 'NEGATIVE': 2}
After: POSITIVE=2, NEGATIVE=2
After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreeshma\Downloads\HTML Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreeshma\Downloads\HTML Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreeshma\Downloads\HTML Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreeshma\Downloads\HTML Tutorials>
```

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.

```
class RecommendationSystem:
    """Product recommendation with fairness and user control"""

    def __init__(self):
        self.user_purchases = {} # {user_id: [products]}
        self.products = {} # {product_id: [category]}
        self.user_opt_out = {} # {user_id: bool}
        self.feedback = {} # {user_id: float}

    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.products[product_id] = {'category': category, 'count': 1}

    def recommend(self, user_id, num=5):
        """
        TRANSPARENCY: show why each recommendation is made
        FAIRNESS: don't only recommend popular products
        USER CONTROL: respect opt-out preferences
        """
        # ETHICAL CHECK: Respect user opt-out
        if user_id in self.user_opt_out:
            return {'status': 'user opted out', 'recommendations': []}

        if user_id not in self.user_purchases:
            return {'status': 'New user', 'recommendations': []}

        user_history = self.user_purchases[user_id]
        user_categories = [self.products[p].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')

                # FAIRNESS: Score based on relevance + diversity
                if category in user_categories:
                    score = 0.5 # Bias toward user's interests
                else:
                    score = 0.5 # Explore new category

                # TRANSPARENCY: Explain why
                reason = f"Similar to your {category} purchases" if category in user_categories else f"Try new: {category}"

                candidates.append({
                    'product': prod_id,
                    'score': score,
                    'reason': reason
                })

        # Sort by score and return top N
        top_recs = sorted(candidates, key=lambda x: x['score'], reverse=True)[:num]

        return {
            'status': 'Success',
            'user_id': user_id,
            'history': user_history,
            'recommendations': top_recs
        }
```

```

13: def recommend(self, user_id, num=2):
14:     """Recommend items for user"""
15:     user_id = int(user_id)
16:     user = self.users[user_id]
17:     user_history = user.history
18:     user_history.sort(key=lambda x: x[1], reverse=True)
19:     return {
20:         "status": "success",
21:         "user_id": user_id,
22:         "history": user_history,
23:         "recommendations": top_recs
24:     }
25:
26: def give_feedback(self, user_id, product_id, liked):
27:     """Collect user feedback to improve fairness"""
28:     self.feedback["user_id"][product_id] = liked
29:     return f"Thanks for feedback on {product_id}"
30:
31: def opt_out(self, user_id):
32:     """Let user opt out of recommendations"""
33:     self.user_opt_out.add(user_id)
34:     return f"{user_id} opted out of recommendations"
35:
36: def opt_in(self, user_id):
37:     """Let user opt back in"""
38:     self.user_opt_out.discard(user_id)
39:     return f"{user_id} opted in to recommendations"
40:
41: # Example Usage
42: if __name__ == "__main__":
43:     system = RecommendationSystem()
44:
45:     # Add purchases
46:     print("=== Adding Purchases ===")
47:     system.add_purchase("user1", "laptop", "Electronics")
48:     system.add_purchase("user1", "mouse", "Electronics")
49:     system.add_purchase("user2", "phone", "Mobile")
50:     print(f"✓ Purchases recorded")
51:
52:     # Add products
53:     system.products["keyboard"] = {"category": "Electronics"}
54:     system.products["monitor"] = {"category": "Electronics"}
55:     system.products["novel"] = {"category": "Books"}
56:
57:     # Get recommendations
58:     print("=== Recommendations for user1 ===")
59:     result = system.recommend("user1", num=2)
60:     for rec in result["recommendations"]:
61:         print(f"Product: {rec['product']}, Score: {rec['score']}, Reason: {rec['reason']}")
62:
63:     # User feedback
64:     print("=== User Feedback ===")
65:     print(system.give_feedback("user1", "keyboard", True))
66:
67:     # Opt out
68:     print("=== User Control ===")
69:     print(system.opt_out("user1"))
70:     result2 = system.recommend("user1")
71:     print(f"✓ After opt-out: {result2['status']}")
72:
73:     # Opt in
74:     print(system.opt_in("user1"))
75:
76:

```

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\jrodriguez\Downloads\WPN_Tutorials> C:\Users\jrodriguez\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\jrodriguez\Documents\AI_ASE\ethical_recommendation_system.py"
--- Adding Purchases ---
PS C:\Users\jrodriguez\Downloads\WPN_Tutorials> C:\Users\jrodriguez\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\jrodriguez\Documents\AI_ASE\ethical_recommendation_system.py"
--- Adding Purchases ---
✓ Purchases recorded

--- Recommendations for user1 ---
Product: keyboard, Score: 0.8, Reason: Similar to your Electronics purchases
Product: monitor, Score: 0.8, Reason: Similar 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 in to recommendations
PS C:\Users\jrodriguez\Downloads\WPN_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.
```

```

1 # See: https://github.com/ethical-security/ethical-security
2 # ethsec:loggy.py
3
4 # Imports
5 import sys
6 import os
7 import logging
8 import re
9
10 # Setup logging
11 class PrivacyFilter(logging.Filter):
12     """Remove sensitive data from logs"""
13
14     def filter(self, record):
15         """Mask passwords, emails, tokens, cards before logging"""
16         msg = str(record.msg)
17
18         # PRIVACY: Mask passwords
19         msg = re.sub(r'password=(.*)', 'password=***REDACTED***', msg, flags=re.IGNORECASE)
20
21         # PRIVACY: Mask emails (show domain only)
22         msg = re.sub(r'([\w\.-]+)@([\w\.-]+)', '[email***]', msg)
23
24         # PRIVACY: Mask auth tokens and hashes
25         msg = re.sub(r'([a-z_]{30})token=[a-z0-9]{40}', '[token=***REDACTED***]', msg, flags=re.IGNORECASE)
26
27         # PRIVACY: Mask credit cards (show last 4 digits)
28         msg = re.sub(r'(\d{4})[0-9]{15}(\d{4})', '\[?(\d{4})[0-9]{15}(\d{4})\]', '[CC****]', msg)
29
30         # PRIVACY: Mask phone numbers (show last 4 digits)
31         msg = re.sub(r'(\d{3})[0-9]{7}(\d{4})', '[PHONE****]', msg)
32
33         record.msg = msg
34         return record
35
36 def setup_logger(name, log_file=None):
37     """Set up logger with privacy protection"""
38     logger = logging.getLogger(name)
39     logger.setLevel(logging.INFO)
40
41     # Add privacy filter
42     privacy_filter = PrivacyFilter()
43
44     # Console handler
45     console_handler = logging.StreamHandler()
46     console_handler.add_filter(privacy_filter)
47     formatter = logging.Formatter('%(asctime)s - %(levelname)s - %(message)s')
48     console_handler.setFormatter(formatter)
49     logger.addHandler(console_handler)
50
51     # File handler
52     if log_file:
53         file_handler = logging.FileHandler(log_file)
54         file_handler.add_filter(privacy_filter)
55         file_handler.setFormatter(formatter)
56         logger.addHandler(file_handler)
57
58     # PrivACY: Restrict file permissions (owner read/write only)
59     import os
60     os.chmod(log_file, 0o600)
61
62     return logger
63
64 def log_user_action(logger, action, user_id, **safe_details):
65     """Log user action with only safe details"""
66     msg = f"{action}: {action} | user: {user_id}"
67     if safe_details:
68         msg += " | " + ", ".join(f"{key}: {value}" for key, value in safe_details.items())
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")

```

[illegible]

Expected Output #4:

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

```
Terminal - VS Code
Test 5: User Action Logging
2020-01-20 10:20:15,606 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}

-----
ETHICAL LOGGING PRACTICES:
-----
1. PRIVACY FILTER: Mask passwords, email, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 600 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVITY: Never store sensitive data in logs
2020-01-20 10:20:15,606 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}

-----
ETHICAL LOGGING PRACTICES:
-----
1. PRIVACY FILTER: Mask passwords, email, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 600 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVITY: Never store sensitive data in logs
2020-01-20 10:20:15,606 - app - INFO - ACTION: purchase | user: user_123 | {'status': 'success', 'amount': 99.99}

-----
ETHICAL LOGGING PRACTICES:
-----
1. PRIVACY FILTER: Mask passwords, email, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 600 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVITY: Never store sensitive data in logs
```

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.

