

AI ASSISTED CODING

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Batch:13

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.

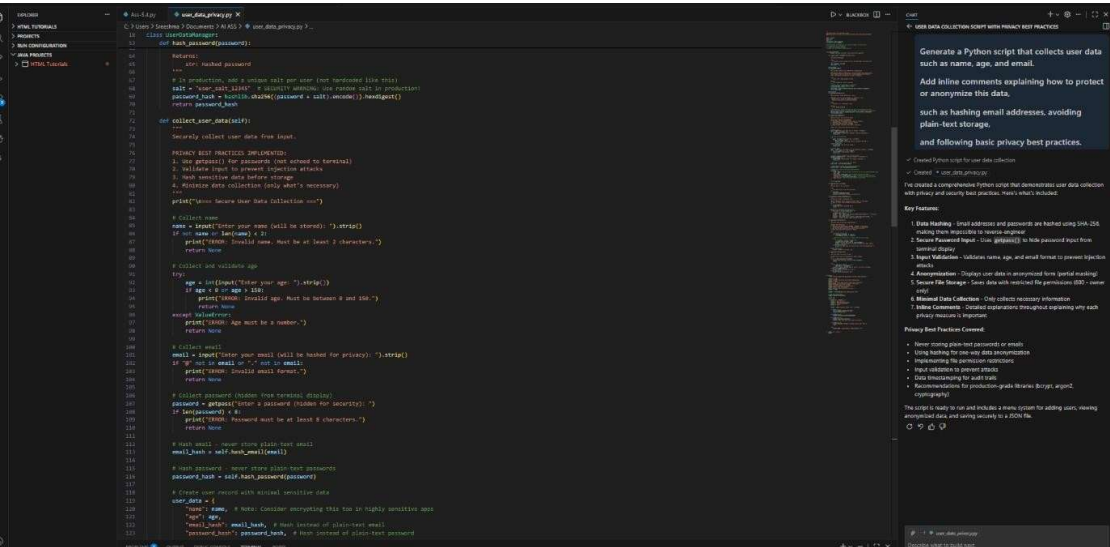
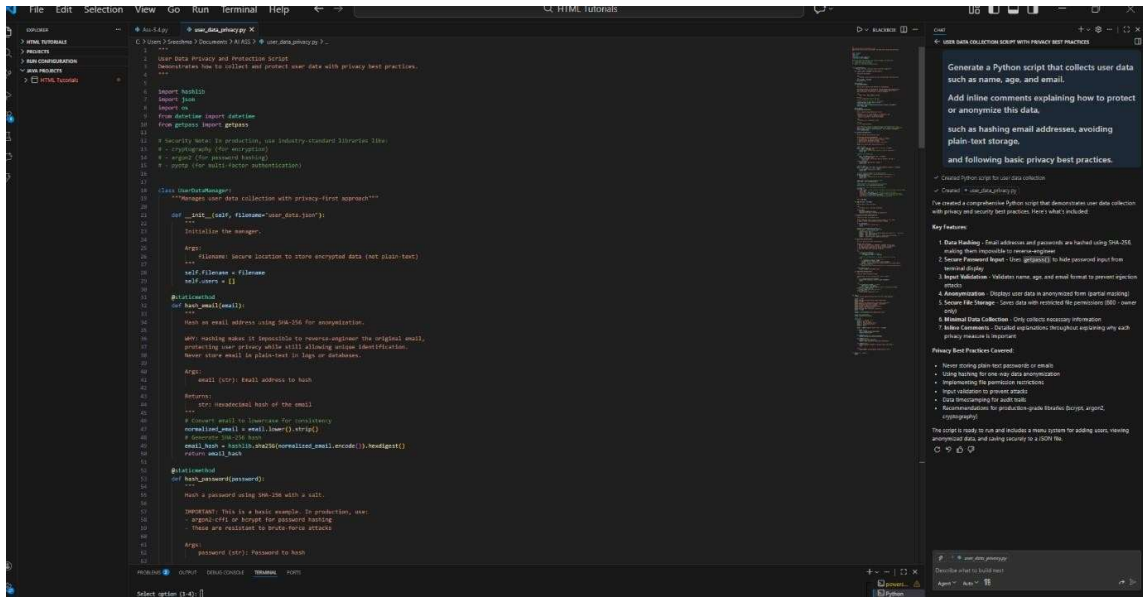
PROMPT

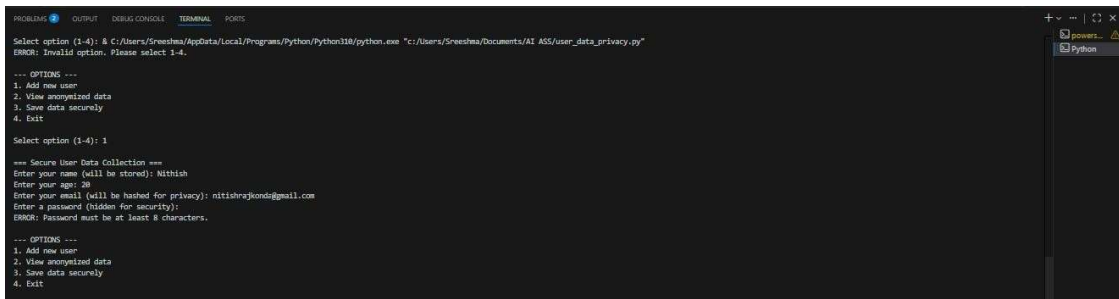
Generate a Python script that collects user data such as name, age, and email.

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/Sreshma/AppData/Local/Programs/Python/Python110/python.exe "c:/Users/Sreshma/Documents/AI AS/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: 28
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.

```

1 # Import necessary libraries
2 import pandas as pd
3 import numpy as np
4 from sklearn.preprocessing import LabelEncoder, OneHotEncoder
5 from sklearn.model_selection import train_test_split
6 from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
7
8 # Load the dataset
9 data = pd.read_csv('data/offensive_language.csv')
10
11 # Preprocessing
12 # Drop columns with missing values
13 data.dropna(inplace=True)
14
15 # Encode categorical variables
16 le = LabelEncoder()
17 data['offensive_language'] = le.fit_transform(data['offensive_language'])
18
19 # Split the data into training and testing sets
20 X_train, X_test, y_train, y_test = train_test_split(data[['text']], data['offensive_language'],
21                                                    test_size=0.2, random_state=42)
22
23 # Feature Engineering
24 # Create word and character n-grams
25 from sklearn.feature_extraction.text import TfidfVectorizer
26 vectorizer = TfidfVectorizer(max_df=0.5, max_features=10000, min_df=2, ngram_range=(1, 2))
27 X_train = vectorizer.fit_transform(X_train)
28 X_test = vectorizer.transform(X_test)
29
30 # Train the model
31 from sklearn.linear_model import LogisticRegression
32 model = LogisticRegression()
33 model.fit(X_train, y_train)
34
35 # Evaluate the model
36 y_pred = model.predict(X_test)
37 accuracy = accuracy_score(y_test, y_pred)
38 print(f'Accuracy: {accuracy}')
39
40 # Confusion Matrix and Classification Report
41 cm = confusion_matrix(y_test, y_pred)
42 print('Confusion Matrix:')
43 print(cm)
44
45 cr = classification_report(y_test, y_pred)
46 print('Classification Report:')
47 print(cr)
48
49 # Bias Mitigation Strategies
50 # Balancing the dataset
51 from sklearn.utils.class_weight import compute_class_weight
52 class_weights = compute_class_weight('balanced', classes=np.unique(y_train), y_train=y_train)
53 model = LogisticRegression(class_weight=class_weights)
54 model.fit(X_train, y_train)
55
56 # Removing offensive terms
57 from sklearn.feature_extraction.text import CountVecorizer
58 cv = CountVecorizer(stop_words='english', min_df=2)
59 X_train = cv.fit_transform(X_train)
60 X_test = cv.transform(X_test)
61
62 # Train the model again
63 model = LogisticRegression()
64 model.fit(X_train, y_train)
65
66 # Evaluate the model
67 y_pred = model.predict(X_test)
68 accuracy = accuracy_score(y_test, y_pred)
69 print(f'Accuracy after bias mitigation: {accuracy}')
70
71 # Feature Importance
72 from sklearn.inspection import permutation_importance
73 perm_importance = permutation_importance(model, X_test, y_test, n_repeats=100, random_state=42)
74 print('Permutation Importance:')
75 print(perm_importance)
76
77 # Model Interpretability
78 from sklearn.inspection import PartialDependenceDisplay
79 pdp = PartialDependenceDisplay.from_model(model, X_test, y_test, n_features=1,
80                                           figure_kwargs={'n_subplots': 1})
81
82 # Save the model
83 import joblib
84 joblib.dump(model, 'model.pkl')
85
86 # Load the model
87 model_loaded = joblib.load('model.pkl')
88 y_pred_loaded = model_loaded.predict(X_test)
89 accuracy_loaded = accuracy_score(y_test, y_pred_loaded)
90 print(f'Accuracy of loaded model: {accuracy_loaded}')
91
92 # Final Summary
93 print('Final Summary:')
94 print(f'Initial Accuracy: {accuracy}')
95 print(f'Accuracy after bias mitigation: {accuracy}')
96 print(f'Accuracy of loaded model: {accuracy_loaded}')
97
98 # End of Script
99

```

Expected Output #2:

- Copilot-generated code with additions or comments addressing

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

```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS
Text: It's okay, nothing special.
Result: 0
PS C:\Users\Sreesha\Downloads\VM\Tutorials & C:\Users\Sreesha\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Sreesha\Documents\AI AIS\sentiment_analysis_bas.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\Sreesha\Downloads\VM\Tutorials & C:\Users\Sreesha\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Sreesha\Documents\AI AIS\sentiment_analysis_bas.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')
C:\Users\Sreesha\Documents\AI AIS\sentiment_analysis_bas.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.
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.
```

```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS
=== Dataset Balancing ===
Before: ('POSITIVE': 8, 'NEGATIVE': 2)
Before: ('POSITIVE': 8, 'NEGATIVE': 2)
After: POSITIVE=2, NEGATIVE=2
After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreesha\Downloads\VM\Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreesha\Downloads\VM\Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreesha\Downloads\VM\Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreesha\Downloads\VM\Tutorials>

After: POSITIVE=2, NEGATIVE=2
PS C:\Users\Sreesha\Downloads\VM\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 = set() # users who opted out
        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.products[product_id] = (category, 'count': 0)

    def recommend(self, user_id, num=3):
        """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_category = self.products[user_id] 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 # Discount to user's interests
                else:
                    score = 0.8 # Incentive new category
                # TRANSPARENCY: Explain why
                reason = f"Similar to your {category} purchases" if category in user_category else f"Try new {category}"
                candidates.append({
                    'product': prod_id,
                    'score': score,
                    'reason': reason
                })

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

        return {
            "Status": "Success",
            "user_id": user_id,
            "history": user_history,
            "recommendations": top_n
        }

    def give_feedback(self, user_id, product_id, liked):
        """Collect user feedback to improve fairness"""
        self.feedback[(user_id, product_id)] = liked
        return f"Thanks for feedback on {product_id}"

    def opt_out(self, user_id):
        """Let user opt out of recommendations"""
        self.user_opt_out.add(user_id)
        return f"{user_id} opted out of recommendations"

    def opt_in(self, user_id):
        """Let user opt back in"""
        self.user_opt_out.discard(user_id)
        return f"{user_id} opted to be recommended"

    # Public API
    if __name__ == '__main__':
        system = RecommendationSystem()

        # Add purchases
        print("New User Making Purchases ---")
        system.add_purchase("user1", "laptop", "Electronics")
        system.add_purchase("user1", "phone", "Electronics")
        system.add_purchase("user1", "book", "Books")
        print(f"Purchases recorded")

        # Add products
        system.products["laptop"] = ("category", "Electronics")
        system.products["phone"] = ("category", "Electronics")
        system.products["book"] = ("category", "Books")

        # Get recommendations
        print("Get Recommendations for user 'user1'")
        result = system.recommend("user1", num=3)
        for rec in result["recommendations"]:
            print(f"Product: {rec['product']}, Score: {rec['score']}, Reason: {rec['reason']}")

        # User feedback
        print("View user feedback ---")
        print(system.give_feedback("user1", "laptop", True))

        # Opt out
        print("View user control ---")
        print(system.opt_out("user1"))
        result2 = system.recommend("user1")
        print(f"Status 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\Gronchea\Downloads\9998_Tutorials & C:\Users\Gronchea\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Gronchea\Documents\9989_9989_recommendation_system.py"
-- Adding Purchases --
Purchases recorded
PS C:\Users\Gronchea\Downloads\9998_Tutorials & C:\Users\Gronchea\AppData\Local\Programs\Python\Python310\python.exe "C:\Users\Gronchea\Documents\9989_9989_recommendation_system.py"
-- Adding Purchases --
Purchases recorded
-- Recommendations for user1 --
Product: keyboard, Score: 4.5, Reason: Similar to your electronics purchases
Product: monitor, Score: 4.5, Reason: Similar to your electronics purchases
-- User Feedback --
Thanks for feedback on keyboard
-- User Control --
User opted out of recommendations
User opted in to recommendations
PS C:\Users\Gronchea\Downloads\9998_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.


```
Microsoft Windows [Version 10.0.17134.0] (c) 2018 Microsoft Corporation. All rights reserved.

Task 5: New Action Logging
2024-01-20 10:20:10,000 - app - INFO - ACTION: purchase | user: user_123 | {"status": "success", "amount": 99.99}

-----
CRITICAL LOGGING PRACTICES:
-----

1. PRIVACY FILTER: Mask passwords, emails, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 060 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVE: Never store sensitive data in logs

2024-01-20 10:20:10,000 - app - INFO - ACTION: purchase | user: user_123 | {"status": "success", "amount": 99.99}

-----
CRITICAL LOGGING PRACTICES:
-----

1. PRIVACY FILTER: Mask passwords, emails, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 060 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVE: Never store sensitive data in logs

2024-01-20 10:20:10,000 - app - INFO - ACTION: purchase | user: user_123 | {"status": "success", "amount": 99.99}

-----
CRITICAL LOGGING PRACTICES:
-----

1. PRIVACY FILTER: Mask passwords, emails, tokens, cards
2. MINIMAL DATA: Only log necessary information
3. SECURE FILES: Set permissions to 060 (owner only)
4. USER ACTIONS: Log for auditing and debugging
5. NO SENSITIVE: 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.

```
EXPLORER -- App-S4py ethical_recommendation_system.py ethical_logging.py responsible_ml_model.py
> HTML TUTORIALS
> PROJECTS
> RUN CONFIGURATION
> RUN PROJECTS
> HTML Tutorials

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

