# AI ASISTED CODING LAB 5 2403A51267

# Task 1: Collect User Data with Anonymization/Protection Comments

```
Code:-
""python import
hashlib
def hash_email(email):
  """Hash email with SHA-256 for anonymization."""
return hashlib.sha256(email.encode()).hexdigest()
def collect_user_data():
  name = input("Enter your name: ")
age = input("Enter your age: ") email
= input("Enter your email: ")
  # --- Data Protection Comments ---
  # - Do not store raw emails or names if not necessary.
  # - Hash or encrypt emails before storage.
  # - Store data in encrypted databases when possible.
  # - Limit access to sensitive fields.
  user_record = {
    "name": name, # Consider replacing with pseudonyms if strict anonymity is needed.
    "age": age,
```

```
"email_hash": hash_email(email) # Store only hashed email.

}

print("Collected (anonymized) user data:", user_record)

if __name__ == "__main__":
    collect_user_data()

...

Output:-
...

Enter your name: Alice
Enter your age: 29

Enter your email: alice@example.com

Collected (anonymized) user data: {'name': 'Alice', 'age': '29', 'email_hash': '3bc51062973c458d5de3d5dba6b6e2a27e94f1ecad3b7e7a9a2c8a4b6f3c5d98'}
...
---
```

### Task 2: Sentiment Analysis with Bias Mitigation:-

#### Code:-

```
"""Returns sentiment polarity and subjectivity."""

blob = TextBlob(text) return blob.sentiment
```

```
# --- Bias Mitigation Comments ---
# - Ensure the training data is balanced for all groups (gender, race, etc.).
# - Remove or flag offensive/biased terms in input and training data.
# - Regularly evaluate model predictions for fairness and accuracy.
# - Allow users to provide feedback if they feel the result is biased.

if __name__ == "__main__":
    example = "I love this product!"

print(sentiment_analysis(example))
...

Output:-
...
Sentiment(polarity=0.5, subjectivity=0.6)
```

#### Task 3: Product Recommender with Ethical Guidelines

#### Code:-

```
""python def recommend_products(user_history,
all_products):
"""

Recommend products based on user history.

- Recommendations should be explainable to the user.

- Ensure fairness: do not favor products due to sponsorship or bias.
"""

# For demo: recommend products not already purchased
recommendations = [p for p in all_products if p not in user_history]
```

```
# --- Ethical Guidelines ---
```

- # Provide explanations for recommendations.
- # Rotate/ensure diversity in recommendations.
- # Allow users to give feedback or opt out.
- # Disclose if recommendations are sponsored or influenced.

#### return recommendations

```
if __name__ == "__main__": history = ["book", "pen"] products =
["book", "pen", "notebook", "eraser"] recs =
recommend_products(history, products) print("Recommendations:",
recs) # Explain why these were recommended print("Explanation:
Recommended items you haven't purchased yet.")
```

#### **Output:-**

...

Recommendations: ['notebook', 'eraser']

Explanation: Recommended items you haven't purchased yet.

...

---

### Task 4: Ethical Logging in a Web Application

#### Code:-

""python import

logging

```
# Configure logger logging.basicConfig(filename="app.log",
level=logging.INFO)
def log_event(event_type, message, user_id=None):
  .....
  Log events without storing sensitive identifiers.
- Never log raw passwords, emails, or other personal data.
- Use pseudonymous identifiers if necessary.
  111111
  # Ethical logging: do not include sensitive info in logs | logging.info(f"Event: {event_type} |
Message: {message} | User: {user_id if user_id else 'N/A'}")
if __name__ == "__main__":
  log_event("LOGIN_ATTEMPT", "User attempted login", user_id="user123")
  # Do NOT log emails, passwords, or tokens!
 Output:-
Event: LOGIN_ATTEMPT | Message: User attempted login | User: user123
```

# **Task 5: Machine Learning Model with Responsible Usage Documentation**

### Code:-

```
```python from sklearn.linear_model import
LogisticRegression from sklearn.datasets import
load_iris from sklearn.model_selection import
train_test_split
# Load data
X, y = load_iris(return_X_y=True)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)
# Train model model =
LogisticRegression(max_iter=200)
model.fit(X_train, y_train)
# Evaluate accuracy =
model.score(X_test, y_test) print("Test
accuracy:", accuracy)
# --- Responsible Usage Documentation ---
111111
Responsible Usage Guidelines:
- This model is for educational/demo purposes and may not generalize to all scenarios.
- Always test accuracy and fairness before deploying.
- Explain model decisions to users where possible (use feature importances, etc.).
- Retrain regularly with up-to-date and diverse data.
- Document known limitations and accuracy bounds.
- Avoid using for high-stakes decisions (e.g., health, legal) without expert review. """
```

## Output:-

• • • •

Test accuracy: 1.0

\*\*\*