

```
In [1]: import os
import numpy as np
import json
import uuid
from datetime import datetime
from pymongo import MongoClient, ASCENDING
import networkx as nx
from itertools import combinations
from collections import defaultdict
from sklearn.metrics.pairwise import cosine_similarity
from flask import Flask, request, jsonify
import mysql.connector
from mysql.connector import Error
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.metrics import confusion_matrix, classification_report, roc_curve, auc, precision_recall_curve
import warnings
warnings.filterwarnings('ignore')

import spacy
nlp = spacy.load("en_core_web_md")
```

----- Vector Database Configuration -----

```
In [2]: DB_CONFIG = {
    'host': 'localhost',
    'user': 'root',
    'password': 'root',
    'database': 'vector_db'
}
```

----- Mongodb Configuration -----

```
In [3]: MONGODB_URI = "mongodb://localhost:27017"
_mongo = MongoClient(MONGODB_URI)
mdb = _mongo["community_db"]
COMM = mdb["communities"]
DB_NAME = "community_db"
COLLECTION = "communities"
```

----- Utility Functions -----

```
In [4]: def keep_nouns_adjs(text):
        doc = nlp(text)
        tokens = [
            token.lemma_.lower()
            for token in doc
            if token.pos_ in ["NOUN", "PROPN", "ADJ"] and not token.is_stop and not token.is_punct
        ]
        return " ".join(tokens)

def text_to_vector(text):
    clean_text = keep_nouns_adjs(text)
    doc = nlp(clean_text)
    return doc.vector
```

----- Vector Database Class -----

```
In [5]: class VectorDatabase:
        def __init__(self, db_config, similarity_threshold=0.6):
            self.similarity_threshold = similarity_threshold
            self.db_config = db_config
            self._init_db()

        def _connect(self):
            return mysql.connector.connect(**self.db_config)

        def _init_db(self):
            conn = self._connect()
            cursor = conn.cursor()
            cursor.execute("""
                CREATE TABLE IF NOT EXISTS vector_database (
                    id VARCHAR(255) PRIMARY KEY,
                    text TEXT NOT NULL,
                    vector TEXT NOT NULL,
                    timestamp VARCHAR(50) NOT NULL
                )
            """)
            conn.commit()
            cursor.close()
            conn.close()
```

```

def add_or_find_duplicate(self, text):
    sim = 0.0
    new_vector = text_to_vector(text)
    conn = self._connect()
    cursor = conn.cursor()
    cursor.execute("SELECT id, text, vector, timestamp FROM vector_database")
    rows = cursor.fetchall()

    for row in rows:
        entry_id, entry_text, entry_vector_json, timestamp = row
        entry_vector = np.array(json.loads(entry_vector_json))
        sim = cosine_similarity(new_vector.reshape(1, -1), entry_vector.reshape(1, -1))[0][0]

        if sim >= self.similarity_threshold:
            cursor.close()
            conn.close()
            return {
                'status': 'duplicate',
                'similarity': sim,
                'id': entry_id,
                'text': entry_text,
                'timestamp': timestamp
            }

    entry_id = str(uuid.uuid4())[:8]
    timestamp = datetime.now().isoformat()
    cursor.execute(
        "INSERT INTO vector_database (id, text, vector, timestamp) VALUES (%s, %s, %s, %s)",
        (entry_id, text, json.dumps(new_vector.tolist()), timestamp)
    )
    conn.commit()
    cursor.close()
    conn.close()

    return {
        'status': 'added',
        'similarity': sim,
        'id': entry_id,
        'text': text,
        'timestamp': timestamp
    }

def calculate_similarity(self, text1, text2):
    """Calculate similarity between two texts without adding to database"""
    vec1 = text_to_vector(text1)
    vec2 = text_to_vector(text2)

```

```

sim = cosine_similarity(vec1.reshape(1, -1), vec2.reshape(1, -1))[0][0]
return sim

def clear_database(self):
    """Clear all entries from the database"""
    conn = self._connect()
    cursor = conn.cursor()
    cursor.execute("DELETE FROM vector_database")
    conn.commit()
    cursor.close()
    conn.close()

```

----- Testing With Synthetic Dataset -----

```

In [6]: df = pd.read_csv(r"C:\Users\Dhineshkumar\Downloads\interest_pairs_100.csv")
df.head()

```

```

Out[6]:

```

	text1	text2	label
0	I love the ocean breeze and sandy shores.	Gardening on weekends helps me connect with na...	0
1	Shooting landscapes and street scenes is my ho...	I love photographing sunsets and everyday life...	1
2	Stories about wizards, dragons, and epic quest...	Running each morning keeps me healthy and focu...	0
3	Exploring fresh songs and curating playlists i...	I enjoy listening to new tracks and sharing pl...	1
4	I like discovering new music and building play...	Finding new artists and organizing playlists m...	1

```

In [7]: similarity_threshold = 0.6
db = VectorDatabase(DB_CONFIG, similarity_threshold=similarity_threshold)
db.clear_database()

```

```

In [8]: predictions = []
true_labels = []
similarity_scores = []

for idx, row in df.iterrows():
    text1 = row['text1']
    text2 = row['text2']
    true_label = int(row['label'])

    # Step 1: clear DB
    db.clear_database()

```

```

# Step 2: insert text1
db.add_or_find_duplicate(text1)

# Step 3: compute similarity by calling the DB method on text2
result = db.add_or_find_duplicate(text2)
sim = result.get('similarity', 0.0)

# Step 4: store results
similarity_scores.append(sim)
pred = 1 if sim >= similarity_threshold else 0
predictions.append(pred)
true_labels.append(true_label)

# Step 5: optional debugging output
if idx < 5:
    print(f"[{idx}]\tttext1[:60]: {text1[:60]!s}")
    print(f"\tttext2[:60]: {text2[:60]!s}")
    print(f"\tsim={sim:.4f} true={true_label} pred={pred}\n")

```

```

[0] text1[:60]: I love the ocean breeze and sandy shores.
    text2[:60]: Gardening on weekends helps me connect with nature.
    sim=0.1100 true=0 pred=0

[1] text1[:60]: Shooting landscapes and street scenes is my hobby.
    text2[:60]: I love photographing sunsets and everyday life in the city.
    sim=0.5112 true=1 pred=0

[2] text1[:60]: Stories about wizards, dragons, and epic quests are my favor
    text2[:60]: Running each morning keeps me healthy and focused.
    sim=0.4271 true=0 pred=0

[3] text1[:60]: Exploring fresh songs and curating playlists is fun for me.
    text2[:60]: I enjoy listening to new tracks and sharing playlists with f
    sim=0.7522 true=1 pred=1

[4] text1[:60]: I like discovering new music and building playlists.
    text2[:60]: Finding new artists and organizing playlists makes my day.
    sim=0.6891 true=1 pred=1

```

----- Metrics -----

```

In [9]: cm = confusion_matrix(true_labels, predictions)
        tn, fp, fn, tp = cm.ravel()

```

```

accuracy = (tp + tn) / (tp + tn + fp + fn)
precision = tp / (tp + fp) if (tp + fp) else 0
recall = tp / (tp + fn) if (tp + fn) else 0
f1_score = 2 * precision * recall / (precision + recall) if (precision + recall) else 0
specificity = tn / (tn + fp) if (tn + fp) else 0

print("\n" + "=" * 70)
print("FINAL EVALUATION RESULTS")
print("=" * 70)
print(f"Accuracy      : {accuracy:.3f}")
print(f"Precision      : {precision:.3f}")
print(f"Recall          : {recall:.3f}")
print(f"F1-Score        : {f1_score:.3f}")
print(f"Specificity     : {specificity:.3f}")
print()
print(f"True Positives : {tp}")
print(f"True Negatives : {tn}")
print(f"False Positives: {fp}")
print(f"False Negatives: {fn}")
print("=" * 70)

```

```

=====
FINAL EVALUATION RESULTS
=====
Accuracy      : 0.760
Precision      : 0.974
Recall         : 0.617
F1-Score       : 0.755
Specificity    : 0.975

True Positives : 37
True Negatives : 39
False Positives: 1
False Negatives: 23
=====

```

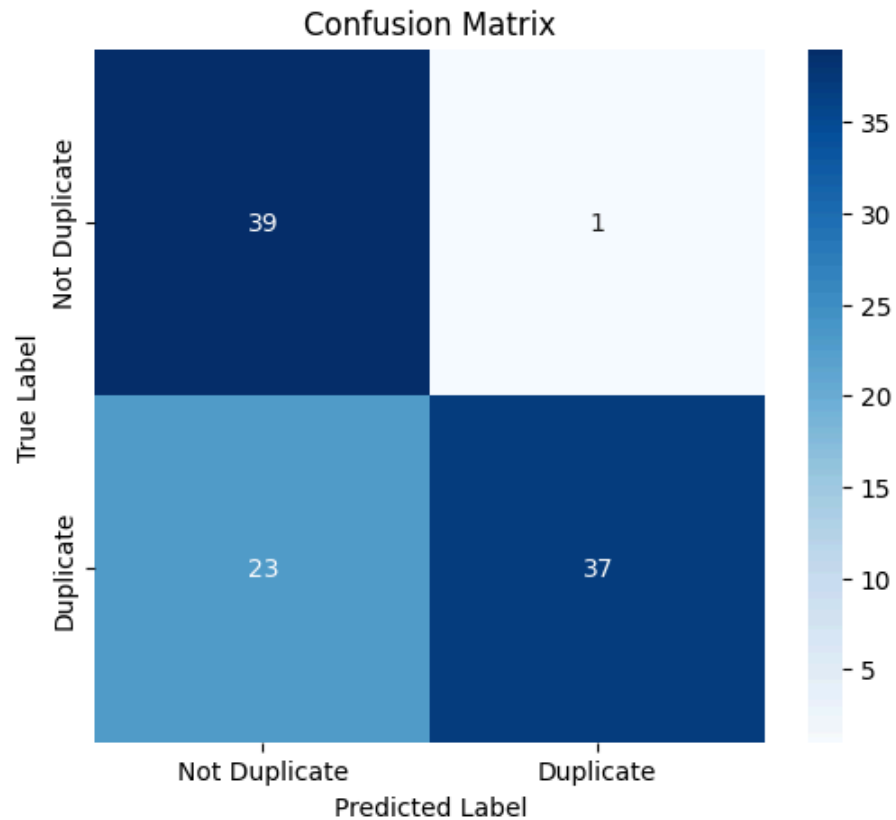
Confusion Matrix

```

In [10]: plt.figure(figsize=(6, 5))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues',
            xticklabels=['Not Duplicate', 'Duplicate'],
            yticklabels=['Not Duplicate', 'Duplicate'])
plt.title('Confusion Matrix')
plt.xlabel('Predicted Label')

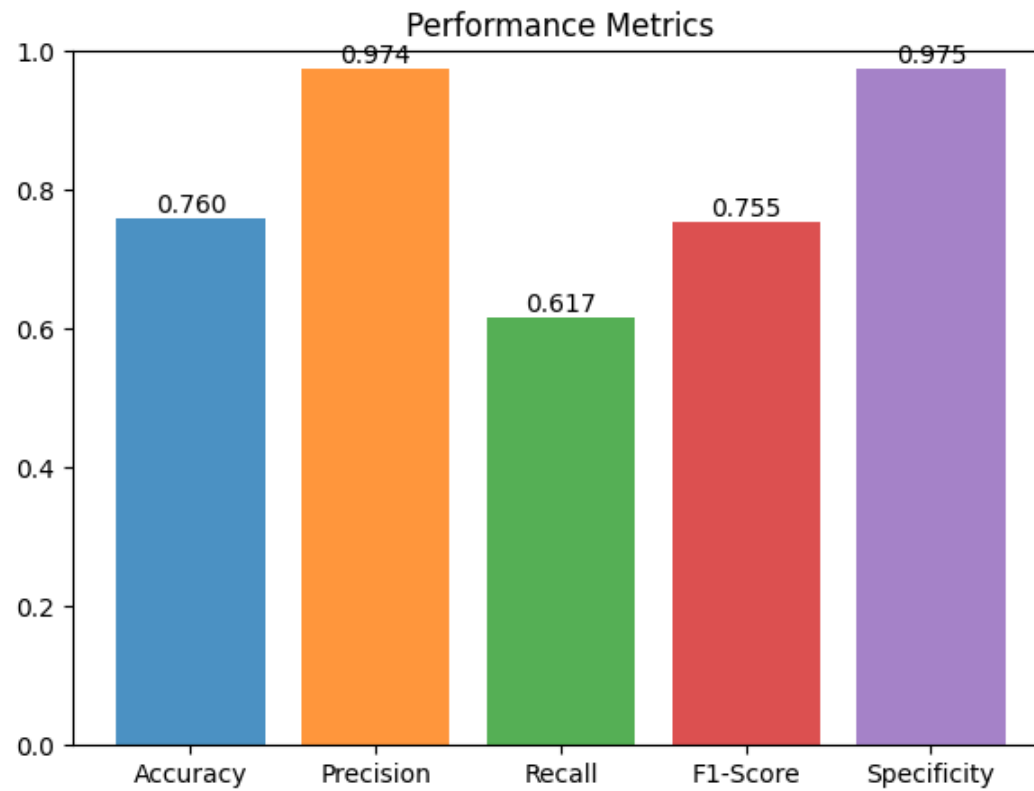
```

```
plt.ylabel('True Label')
plt.show()
```



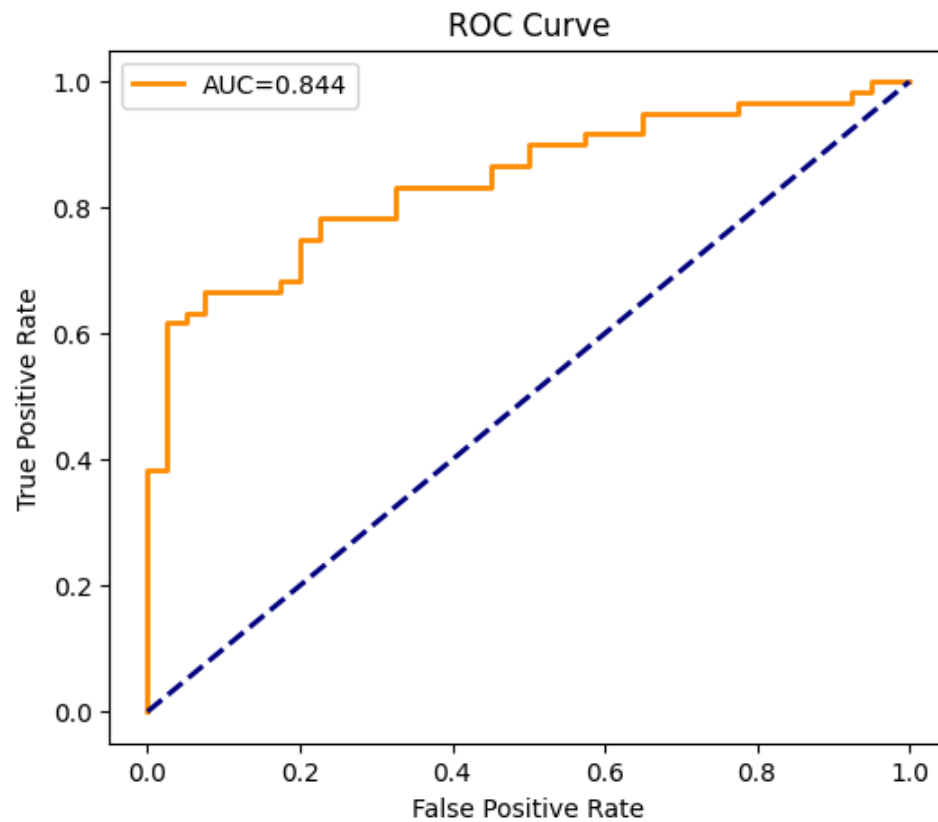
Bar Chart

```
In [11]: plt.figure(figsize=(7, 5))
metrics = ['Accuracy', 'Precision', 'Recall', 'F1-Score', 'Specificity']
values = [accuracy, precision, recall, f1_score, specificity]
colors = ['#1f77b4', '#ff7f0e', '#2ca02c', '#d62728', '#9467bd']
bars = plt.bar(metrics, values, color=colors, alpha=0.8)
plt.ylim(0, 1)
plt.title('Performance Metrics')
for bar in bars:
    plt.text(bar.get_x() + bar.get_width()/2, bar.get_height(),
             f"{bar.get_height():.3f}", ha='center', va='bottom')
plt.show()
```



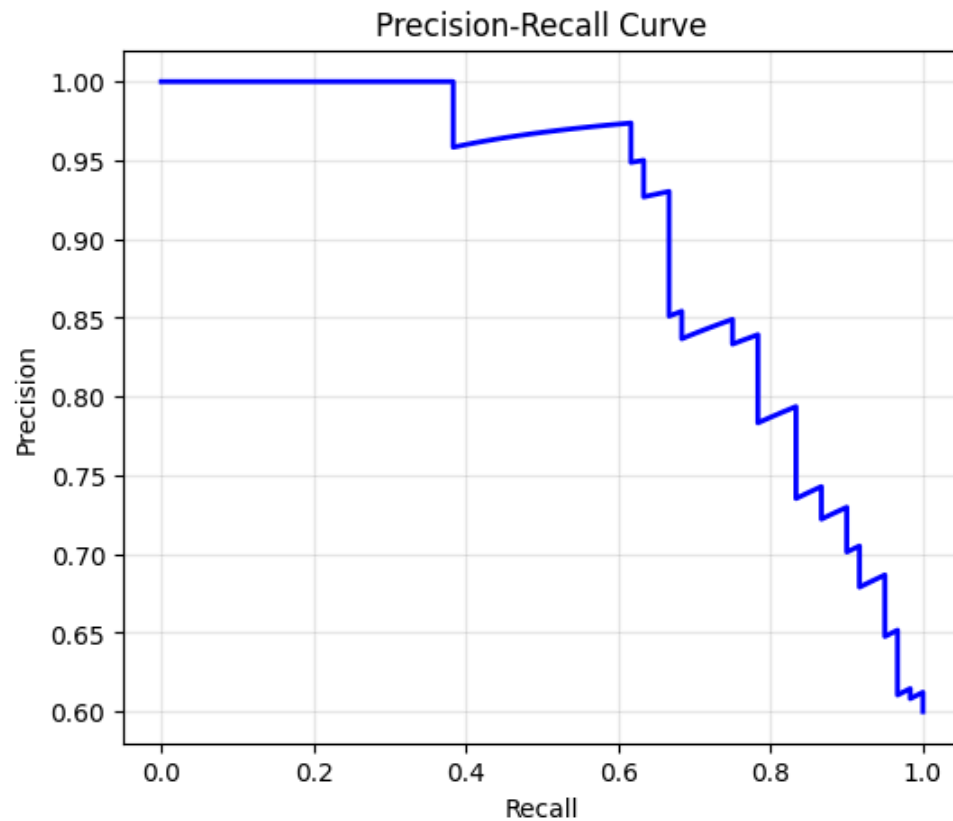
ROC Curve

```
In [12]: plt.figure(figsize=(6, 5))
fpr, tpr, _ = roc_curve(true_labels, similarity_scores)
roc_auc = auc(fpr, tpr)
plt.plot(fpr, tpr, color='darkorange', lw=2, label=f'AUC={roc_auc:.3f}')
plt.plot([0, 1], [0, 1], color='navy', lw=2, linestyle='--')
plt.legend()
plt.title('ROC Curve')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.show()
```

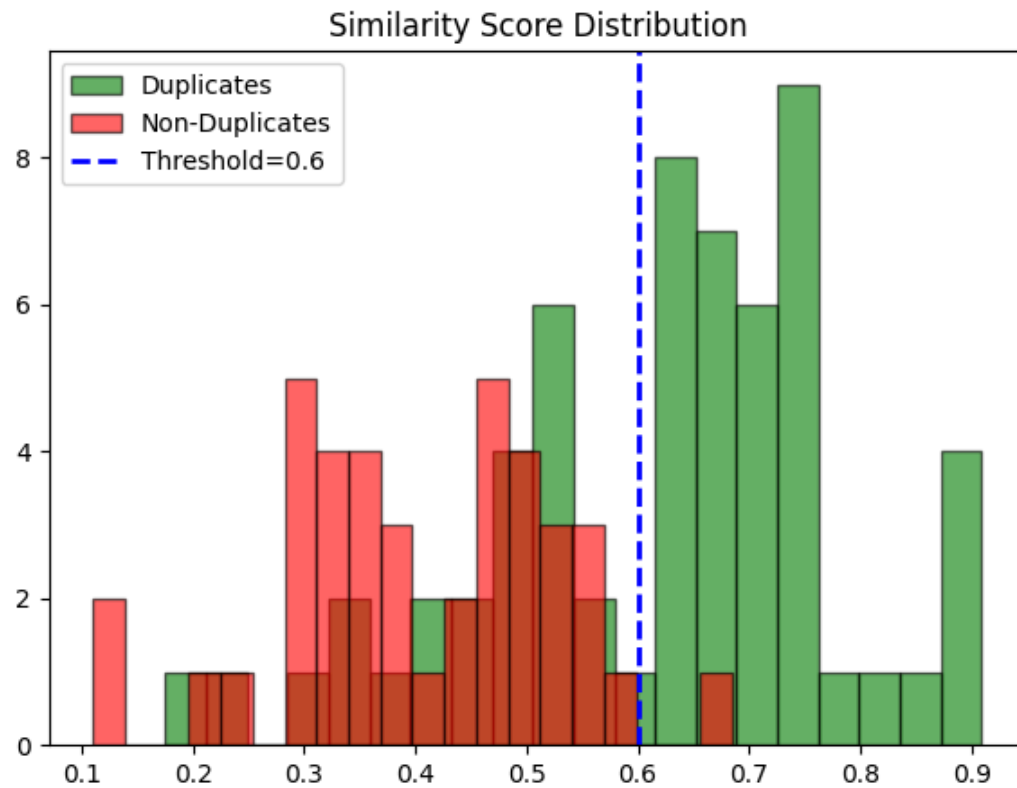
Precision-Recall Curve

```
In [13]: plt.figure(figsize=(6, 5))
precision_curve, recall_curve, _ = precision_recall_curve(true_labels, similarity_scores)
plt.plot(recall_curve, precision_curve, color='blue', lw=2)
plt.title('Precision-Recall Curve')
plt.xlabel('Recall')
plt.ylabel('Precision')
plt.grid(alpha=0.3)
plt.show()
```



Similarity Distribution

```
In [14]: plt.figure(figsize=(7, 5))
duplicate_scores = [similarity_scores[i] for i in range(len(similarity_scores)) if true_labels[i] == 1]
non_duplicate_scores = [similarity_scores[i] for i in range(len(similarity_scores)) if true_labels[i] == 0]
plt.hist(duplicate_scores, bins=20, alpha=0.6, label='Duplicates', color='green', edgecolor='black')
plt.hist(non_duplicate_scores, bins=20, alpha=0.6, label='Non-Duplicates', color='red', edgecolor='black')
plt.axvline(similarity_threshold, color='blue', linestyle='--', linewidth=2, label=f'Threshold={similarity_threshold}')
plt.title('Similarity Score Distribution')
plt.legend()
plt.show()
```



----- SNA Metrics -----

```
In [15]: # stack text1 and text2 into a single column
df_flat = pd.DataFrame({
    "text": pd.concat([df["text1"], df["text2"]], ignore_index=True)
})

df_flat["user_id"] = [str(uuid.uuid4()) for _ in range(len(df_flat))]
```

```
In [16]: df_flat.head()
```

Out[16]:

		text	user_id
0		I love the ocean breeze and sandy shores.	cbe01cf3-1459-4cf5-ab60-20eaa6d3bb1e
1		Shooting landscapes and street scenes is my ho...	dc05f62a-9cd5-4cbe-b4e6-269f669f4fe2
2		Stories about wizards, dragons, and epic quest...	105980f6-db29-4725-9927-b42a61e17518
3		Exploring fresh songs and curating playlists i...	35a343c6-c274-4c36-ade3-0a7ccb52970c
4		I like discovering new music and building play...	5258eb6c-251f-4440-9006-fb7673dcea94

SNA Metrics for threshold = 0.4

```
In [17]: db = VectorDatabase(DB_CONFIG, similarity_threshold=0.4)
```

```
In [18]: #giving user id
df_flat["user_id"] = [str(uuid.uuid4()) for _ in range(len(df_flat))]

def now_iso():
    return datetime.utcnow().isoformat()

# build communities
for txt, uid in zip(df_flat["text"], df_flat["user_id"]):
    out = db.add_or_find_duplicate(txt)
    group_id = out["id"]

    COMM.update_one(
        {"_id": group_id},
        {
            "$setOnInsert": {
                "created_at": now_iso(),
                "threshold": db.similarity_threshold
            },
            "$addToSet": { "users": uid }
        },
        upsert=True
    )
```

```
In [19]: # Build user-user graph (edge if two users appear in the same community)
def build_graph_from_mongo(uri=MONGODB_URI, db=DB_NAME, coll=COLLECTION):
    client = MongoClient(uri)
    COMM = client[db][coll]
    G = nx.Graph()
```

```

edge_w = defaultdict(int)
nodes = set()

for doc in COMM.find({}, {"users": 1}):
    users = [u for u in (doc.get("users") or []) if u is not None]
    for u in users:
        nodes.add(u)
    for a, b in combinations(sorted(set(users)), 2):
        edge_w[(a, b)] += 1

G.add_nodes_from(nodes)
for (a, b), w in edge_w.items():
    G.add_edge(a, b, weight=w)
return G

G = build_graph_from_mongo()
print(f"Graph built: {G.number_of_nodes()} nodes, {G.number_of_edges()} edges")

```

Graph built: 200 nodes, 5084 edges

Graph Density

```

In [20]: density = nx.density(G) if G.number_of_nodes() > 1 else 0.0
print(f"Graph Density: {density:.3f}")

```

Graph Density: 0.255

Average Clustering Coefficient

```

In [21]: avg_clustering = nx.average_clustering(G) if G.number_of_nodes() > 0 else 0.0
print(f"Average Clustering Coefficient: {avg_clustering:.3f}")

```

Average Clustering Coefficient: 0.990

Degree Centrality

```

In [22]: degree centrality = nx.degree_centrality(G) if G.number_of_nodes() > 0 else {}
max_degree = max(degree_centrality.values()) if degree_centrality else 0.0
print(f"Highest Degree Centrality Score: {max_degree}")

```

Highest Degree Centrality Score: 0.4221105527638191

Closeness Centrality

```
In [23]: closeness centrality = nx.closeness centrality(G) if G.number_of_nodes() > 0 else {}  
max_closeness = max(closeness centrality.values()) if closeness centrality else 0.0  
print(f"Highest Closeness Centrality Score: {max_closeness}")
```

Highest Closeness Centrality Score: 0.4221105527638191

Betweenness Centrality

```
In [24]: betweenness centrality = nx.betweenness centrality(G, normalized=True) if G.number_of_nodes() > 0 else {}  
max_betweenness = max(betweenness centrality.values()) if betweenness centrality else 0.0  
print(f"Highest Betweenness Centrality Score: {max_betweenness}")
```

Highest Betweenness Centrality Score: 0.0

Users Belonging to more than one community

```
In [25]: user_to_comms = defaultdict(list)  
  
for doc in COMM.find({}, {"_id": 1, "users": 1}):  
    community_id = doc["_id"]  
    users = doc.get("users", [])  
    for u in users:  
        user_to_comms[u].append(community_id)  
  
multi_community_users = {u: comms for u, comms in user_to_comms.items() if len(comms) > 1}  
  
print("== Users in more than one community ==")  
if multi_community_users:  
    for user, comms in multi_community_users.items():  
        print(f"User: {user}\n Communities: {comms}\n")  
else:  
    print("No users found in more than one community.")
```

== Users in more than one community ==
No users found in more than one community.