

Prompts for assignment 2

Q1

1. Learning rate issues

<https://chatgpt.com/share/68c85254-6114-8013-9040-8e50edd1b0c5>

2. Fixes - <https://g.co/gemini/share/52c04a1a5049>

Q3

Initialization and categorical variable doubts -

Implement the GMM Class [6 marks]

- Write your own GMM class.
- Your GMM class should include methods like fit(), getMembership() and getLikelihood().
- The fit() method implements the Expectation-Maximization (EM) algorithm on the dataset to determine the optimal parameters for the model.

in gmm how does one do initialization for this fit method currently have made the init only
should I initialize it to random ??

```
class GMM:  
  
    def __init__(self, n_clusters=3):  
        self.n_clusters = n_clusters  
        self.means = []  
        self.covariances = []  
        self.mix_coeff = []
```

???

Here's how you could extend your `__init__` and handle initialization in `fit()`:

```
python Copy code  
  
import numpy as np  
  
class GMM:  
    def __init__(self, n_clusters=3, max_iter=100, tol=1e-6, random_state=None):  
        self.n_clusters = n_clusters  
        self.max_iter = max_iter  
        self.tol = tol  
        self.random_state = np.random.RandomState(random_state)  
  
        # Parameters to learn  
        self.means = None  
        self.covariances = None  
        self.mix_coeff = None  
  
    def _initialize_parameters(self, X):  
        n_samples, n_features = X.shape  
  
        # Initialize means by randomly choosing points  
        random_idx = self.random_state.choice(n_samples, self.n_clusters, replace=False)  
        self.means = X[random_idx]  
  
        # Initialize covariances as identity matrices  
        self.covariances = [np.cov(X.T) for _ in range(self.n_clusters)]  
  
        # Initialize mixing coefficients uniformly  
        self.mix_coeff = np.ones(self.n_clusters) / self.n_clusters  
  
    def fit(self, X):  
        self._initialize_parameters(X)  
        # then run EM loop:  
        # E-step: compute responsibilities  
        # M-step: update means, covariances, mix_coeff  
        pass  
  
    def getMembership(self, X):  
        # Return responsibilities (soft assignments)  
        pass  
  
    def getLikelihood(self, X):  
        # Compute log-likelihood of data  
        pass
```

CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)	Gender
0	Male	19.0	15.0	39.0	1
1	Male	21.0	15.0	81.0	1
2	Female	20.0	16.0	6.0	0
3	Female	23.0	16.0	77.0	0
4	Female	31.0	17.0	40.0	0

DF is like this so I will be using the genre

in gmm we dont do normalization or stuff or we do ???

Like I will not use customer id as a feature rest all will be used

• Which features to use?

- Use: Age, Annual Income, Spending Score.
- Gender: You need to be careful.
 - GMM assumes **continuous Gaussian features**. Encoding Gender as `0/1` makes it look like a numeric variable, but it's actually categorical.
 - If you include it, the Gaussian assumption is a bit wrong — but sometimes still used in practice.
Alternatively, you can:
 - Drop it if it's not useful for segmentation.
 - Or treat it separately (e.g., fit a separate GMM per gender).
 - Do not use `CustomerID` → just an identifier.

• Workflow for your case

```
python Copy code  
  
import pandas as pd  
from sklearn.preprocessing import StandardScaler  
  
# Load your dataframe (df)  
  
# Drop CustomerID  
X = df.drop(columns=["CustomerID", "Genre"]) # or keep Gender if you want  
  
# Standardize  
scaler = StandardScaler()  
X_scaled = scaler.fit_transform(X)
```

Then you can pass `X_scaled` into your `GMM.fit()`.

Multivariate gaussian - <https://g.co/gemini/share/075c71a3897d> later final fixes done with copilot -

I understand. Let me first get the current notebook summary and then check the multivariate normal class to make minimal changes only to fix the issue without modifying the pdf function logic.

✓ Retrieving Notebook summary.

Now let me check the current multivariate normal class:

✓ Read q3.ipynb , lines 40 to 90

Perfect! I can see the issue. The problem is in line 41 where it uses `axis=1` but when a single sample is passed (1D array), there is no axis=1. The fix is simple - I just need to handle the input `x` to ensure it's always 2D, even for single samples. This way I don't change the pdf function logic, just ensure the input is in the right shape.

Now let's test the fix by running the problematic cell:

Q4

1. Some bugs and help -

<https://chatgpt.com/share/68d6f070-dc68-8001-ab24-8bcfcff93a11>

2. Video making with claude (copilot)

in this I have to make the video in each frame I will map the image we on left always the original image then in mid have to show the segmented image and have to grow the likelihood graph in the last image so have to make a video of each frame say lasting 0.75 seconds so how can i make the video just change the last cell accordingly do import there if needed

I'll help you modify the last cell to create a video with three panels: original image, segmented image, and growing likelihood graph. Let me first examine the current notebook to understand the structure.

✓ Retrieving Notebook summary.

Now let me read the current last cell to understand the existing code structure:

✓ Read q4.ipynb , lines 250 to 272

Now let me read more context to understand the existing GMM implementation and helper functions:

✓ Read q4.ipynb , lines 150 to 200

```
import numpy as np
def create_video_with_panels(img, output_filename, frame_duration=0.75):
    """
    Create a video with three panels:
    - Left: Original image
    - Middle: Segmented image (evolving)
    - Right: Likelihood graph (growing)
    """
    # Get segmented images and likelihoods
    df = convert_to_df(img)
    x = df[['R', 'G', 'B']]
    gmm_obj = GMM(x, n_clusters=3, max_iters=10)
    gmm_obj.fit()

    all_time_resp = gmm_obj.resp.all_time
    likelihoods = gmm_obj.likelihood

    # Generate segmented images for each iteration
    segmented_images = []
    for resp in all_time_resp:
        seg_img = get_segmented_image(img, resp, n_clusters=3)
        segmented_images.append(seg_img)

    # Convert original image to numpy array
    original_img = np.array(img)

    # Set up the figure with three subplots
    fig, (ax1, ax2, ax3) = plt.subplots(1, 3, figsize=(15, 5))
    fig.suptitle('GMM Image Segmentation Evolution', fontsize=16)

    # Display original image (static)
    ax1.imshow(original_img)
    ax1.set_title('Original Image')
    ax1.axis('off')

    # Initialize middle panel for segmented image
    ax2.set_title('Segmented Image')
    ax2.axis('off')
    seg_display = ax2.imshow(segmented_images[0])

    # Initialize right panel for likelihood plot
    ax3.set_title('Log-Likelihood Evolution')
    ax3.set_xlabel('Iteration')
    ax3.set_ylabel('Log-Likelihood')
    time_ = np.arange(0, len(all_time_resp) - 1)

    # Animate the likelihood graph
    def update(frame):
        seg_display.set_data(segmented_images[frame])
        ax3.clear()
        ax3.plot(time_, all_time_resp[:, frame], 'r')
        ax3.set_xlim(0, len(all_time_resp) - 1)
        ax3.set_ylim(0, 1.0)

    # Create the animation
    anim = FuncAnimation(fig, update, frames=len(all_time_resp) - 1, interval=frame_duration)
    anim.save(output_filename, fps=10)
```

Better visualization for this -

<https://g.co/gemini/share/e0d51758f7e7>

Q5

1. Understanding basics and issues -

<https://chatgpt.com/share/68d79fb6-5328-8001-af1e-e7c59b1d2ef0>

Q7

1. Getting the code for the logic I had - <https://q.co/gemini/share/05558da52f1b>
<https://chatgpt.com/share/68d818ab-8780-800f-90f9-92fe9f336d0c>