Step 3 – Fuzzy C-means Application

1. Apply the **Fuzzy C-means** algorithm (use **c = 2**).
2. Analyze the **membership degrees** of each point to each cluster.
3. Plot the points by **color intensity** according to the membership to **Class 1** (e.g., higher intensity → higher membership).

⚙️ When to Use Fuzzy C-means (FCM)

- ✅ Use it when:
  - You expect **overlapping clusters** and want **soft assignments** (degrees of belonging).
  - Boundaries between groups are **gradual**, not crisp (e.g., transitions or mixtures).
  - You need **membership values** for downstream tasks (weights, uncertainty, interpolation).
  - Data is **continuous** and distance-based separation is meaningful.
- 🚫 Avoid it when:
  - You require **hard labels only** and overlaps are negligible (K-means may suffice).
  - Data has **strong outliers** or very **non-spherical** shapes (consider DBSCAN/OPTICS).
  - Features are **categorical** without a proper numeric embedding.
  - You can't afford iterative membership updates on **very large datasets**.

🤖 Model Hyperparameters

- `c = 2` — number of clusters
- `m = 2.0` — fuzziness coefficient (higher → softer membership)
- `max_iter = 200` — maximum number of iterations
- `tol = 1e-5` — convergence threshold
- `seed = 42` — random initialization seed

```
%run 00-setup.py
```

```
from ml.viz import plt_fcm
from tasks.fcm import run_fcm
from ml.data import load_dataset
```

```
X, y, _ = load_dataset("../data/data_bivariate_gaussian.npz")
```

```
res = run_fcm(X, c=2, m=2.0, max_iter=200, tol=1e-5, seed=42)

U = res["U"]
labels = res["labels"]
```

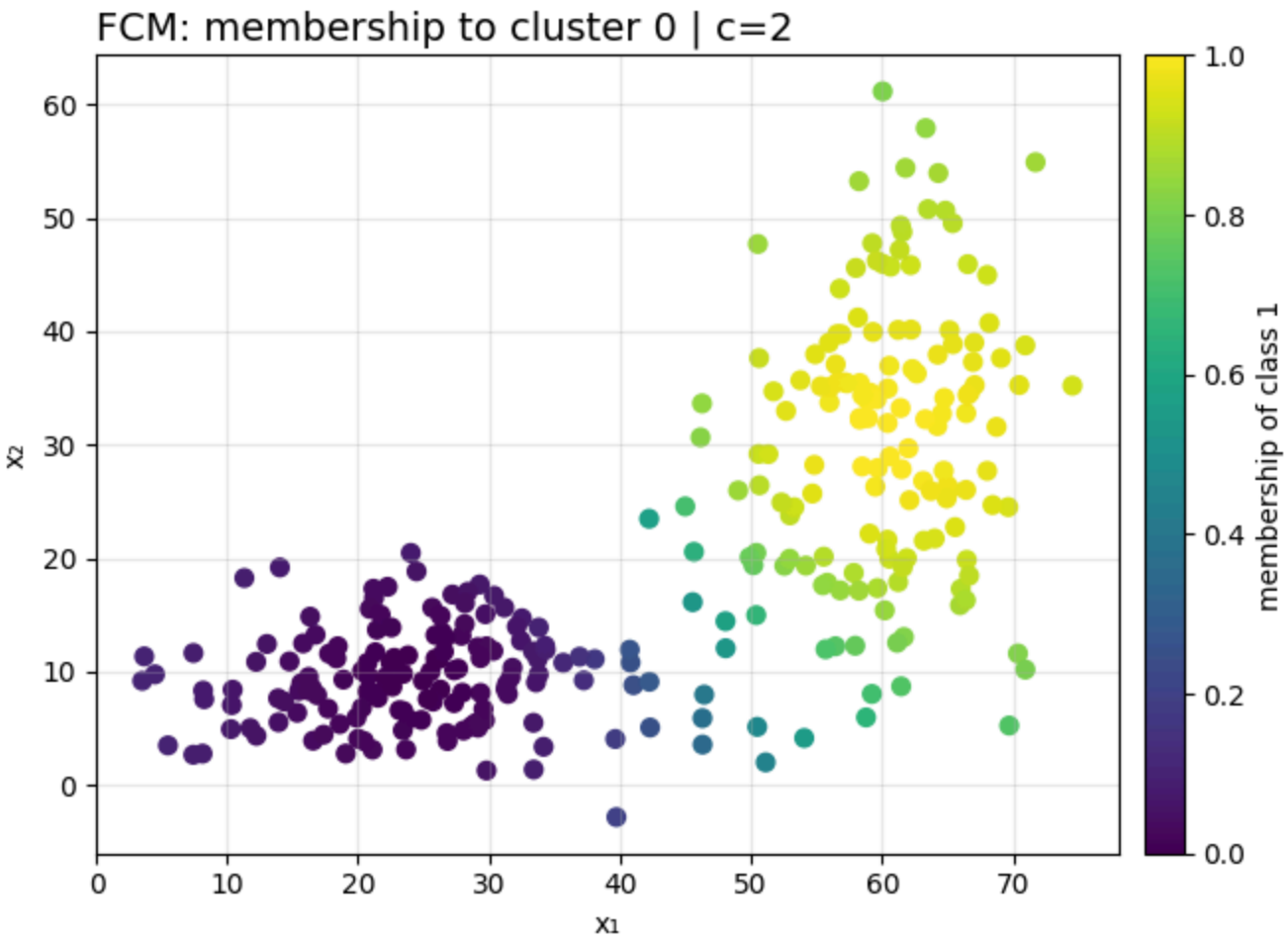
```
centers = res["centers"]

print("Centers:\n", centers, "\n")
print("Iterations:", res["n_iter"])
```

```
Centers:
 [[60.07332666 30.88583435]
 [24.5111101   9.79764972]]

Iterations: 15
```

```
plt_fcm(X, U[0], title="FCM: membership to cluster 0 | c=2")
```




```
(<Figure size 700x500 with 2 Axes>,
 <Axes: title={'left': 'FCM: membership to cluster 0 | c=2'}, xlabel='x₁', ylabel='x₂'>)
```