

# CAM-SOFTMAX FOR DISCRIMINATIVE DEEP FEATURE LEARNING

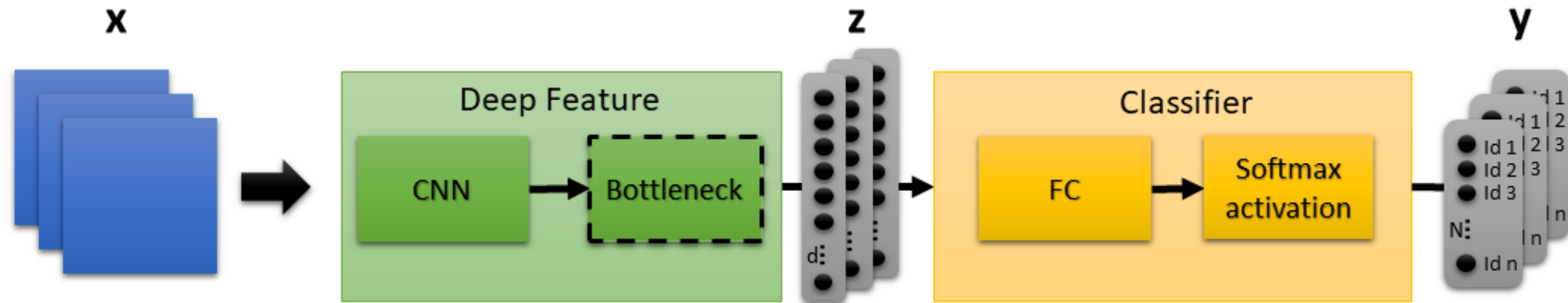
Tamás Süveges

Stephen McKenna

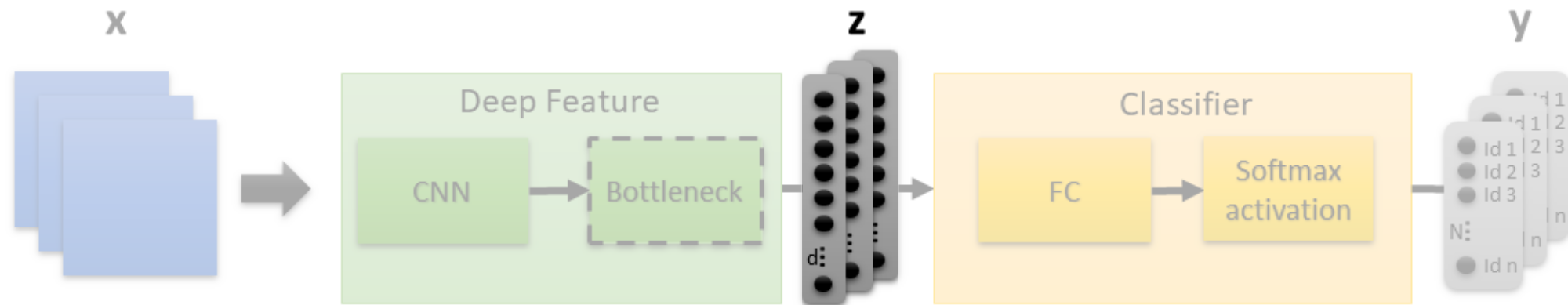
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Research Council (EPSRC grant EP/N014278/1)

# Deep learning classification pipeline

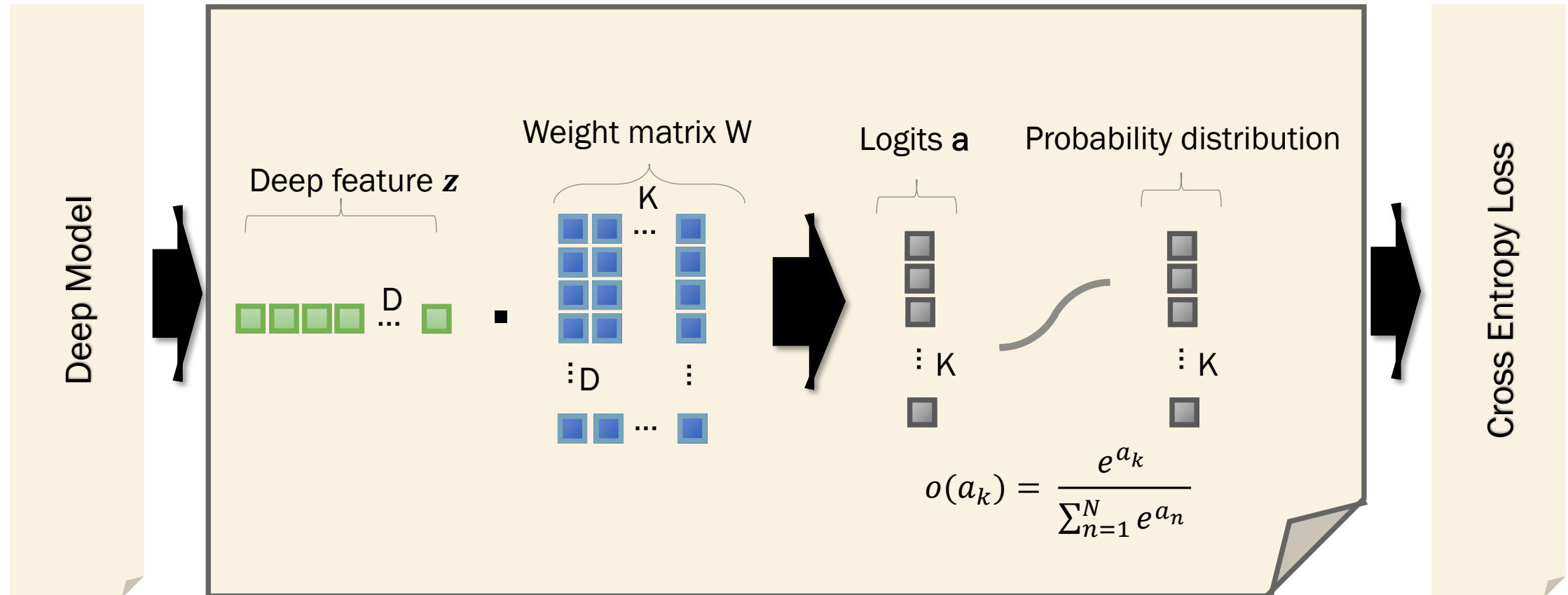


# Deep learning classification pipeline



- ✓ Good for out-of-set recognition
- ✓ Graceful degradation in the presence of label noise

# Terminating layer



# Generalized softmax

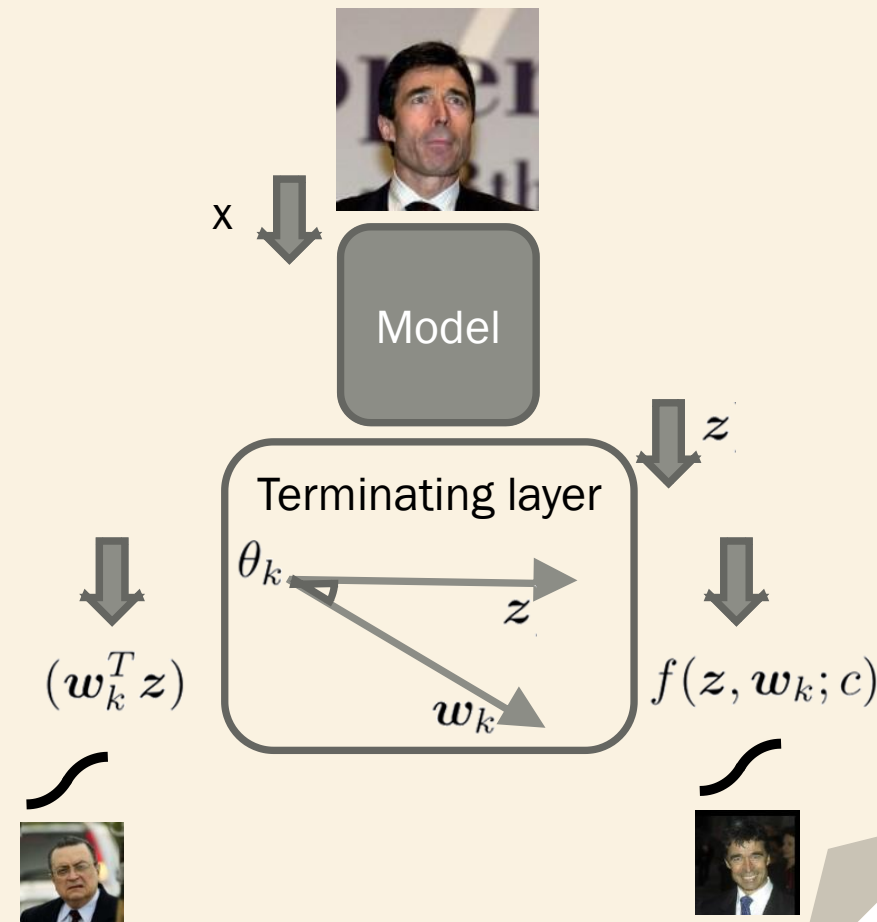
*cam – softmax*

$$y_k = \frac{e^{a_k}}{\sum_j e^{a_j}}$$

$$a_k = \mathbf{1}(k \neq t)(\mathbf{w}_k^T \mathbf{z}) + \mathbf{1}(k = t)(f(\mathbf{z}, \mathbf{w}_k; c))$$

$$f(\mathbf{z}, \mathbf{w}_k; c) = \frac{(\cos \theta_k + 1)^{g(c)}}{2^{(g(c)-1)}} - 1$$

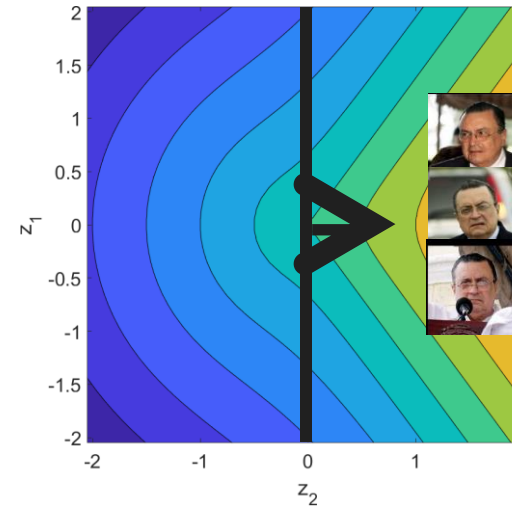
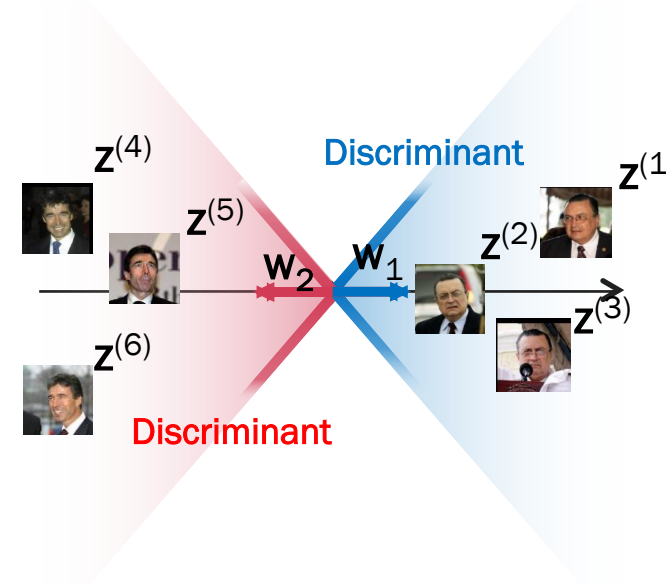
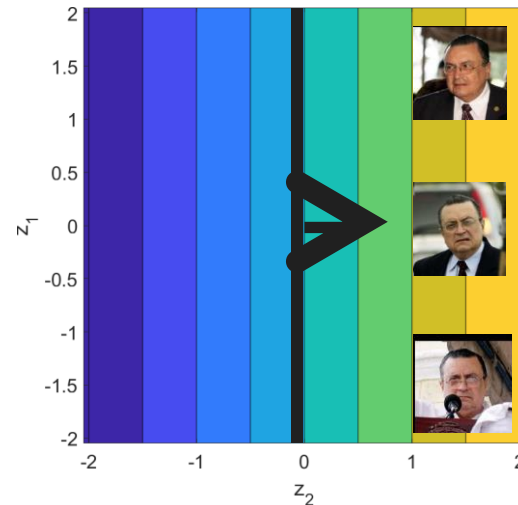
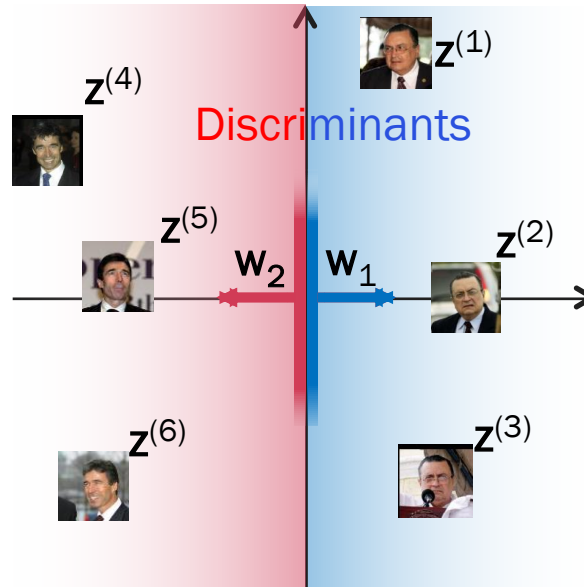
*Intuitive illustration*



# Geometrical intuition

Softmax

Unnecessarily spread

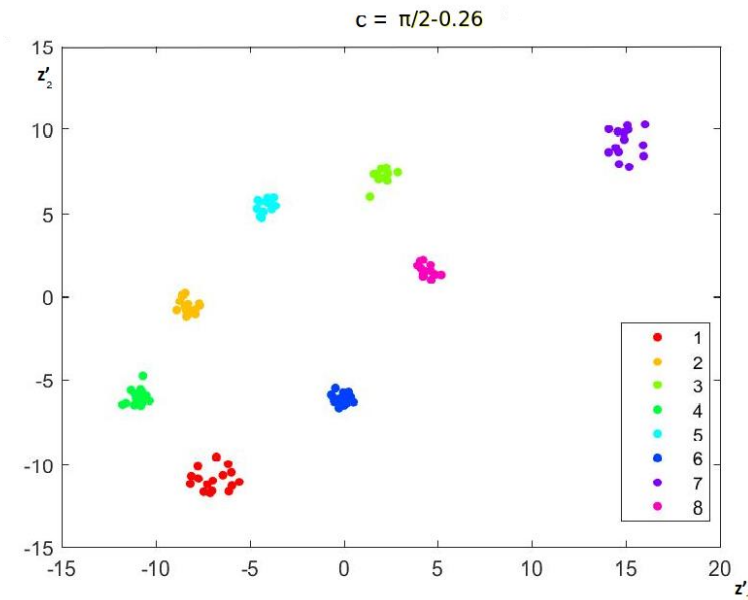
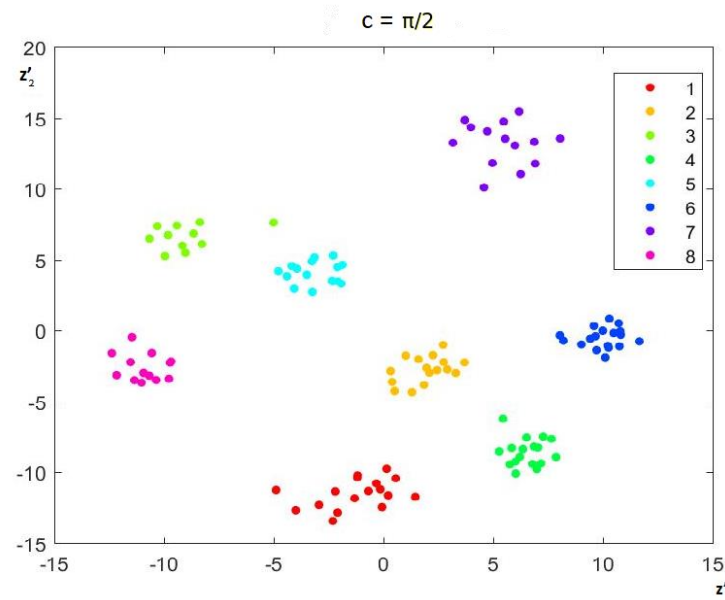


Well concentrated

Cam-softmax

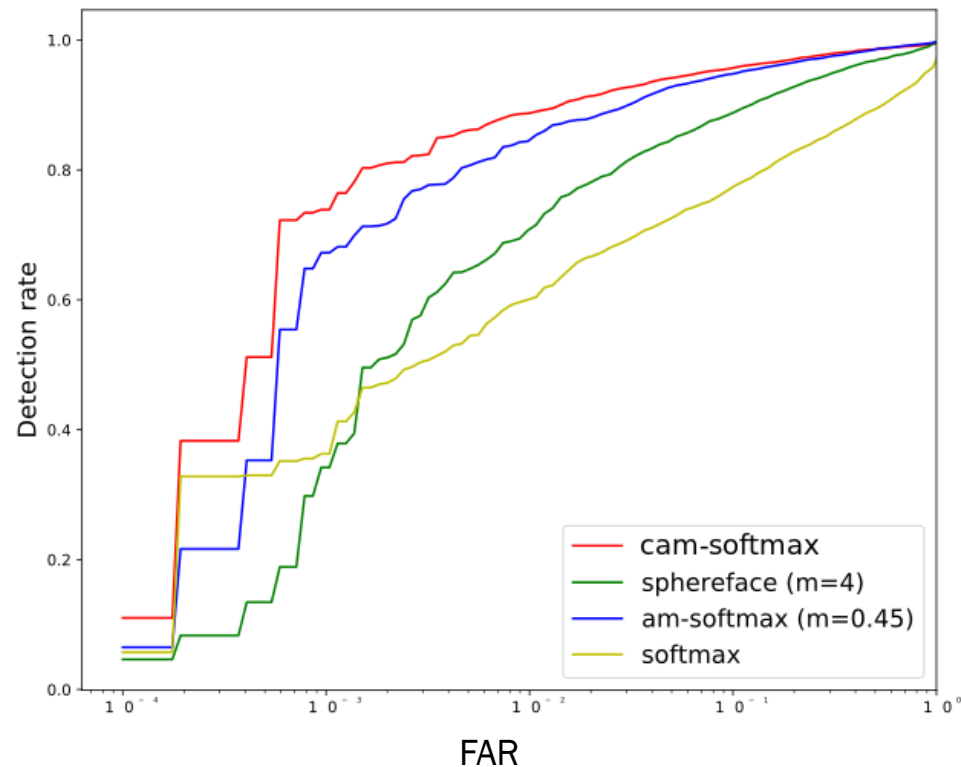
# Visualization with t-SNE

8 random identities from the Labelled  
Faces in the Wild (LFW) dataset

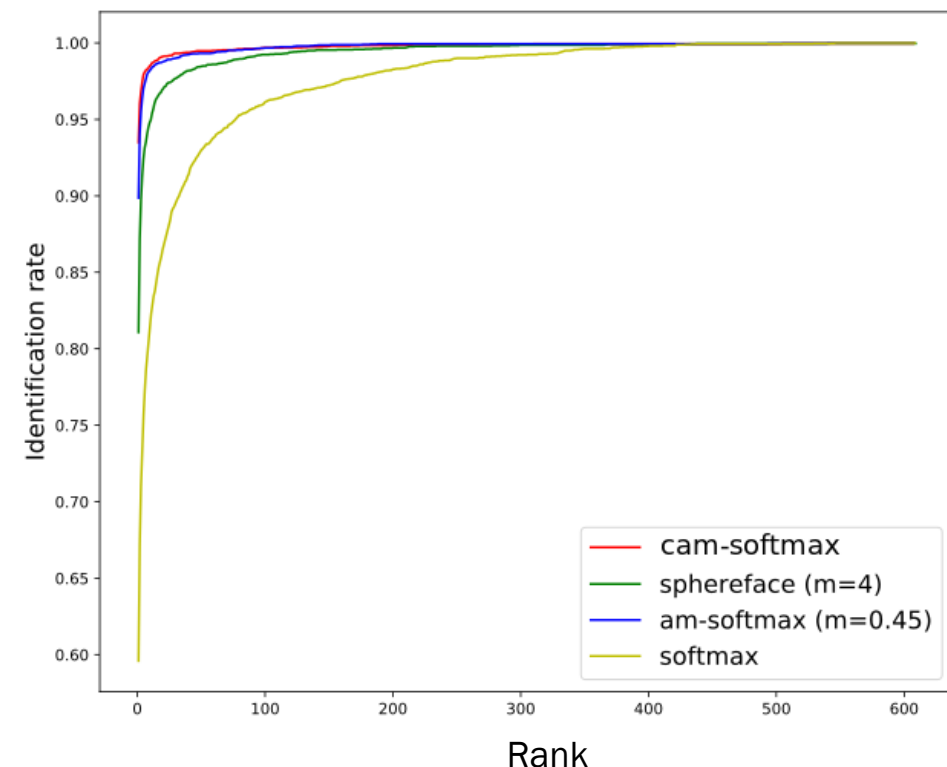


# Open-set Recognition On LFW

How well the model can differentiate between known identities and the unknowns?

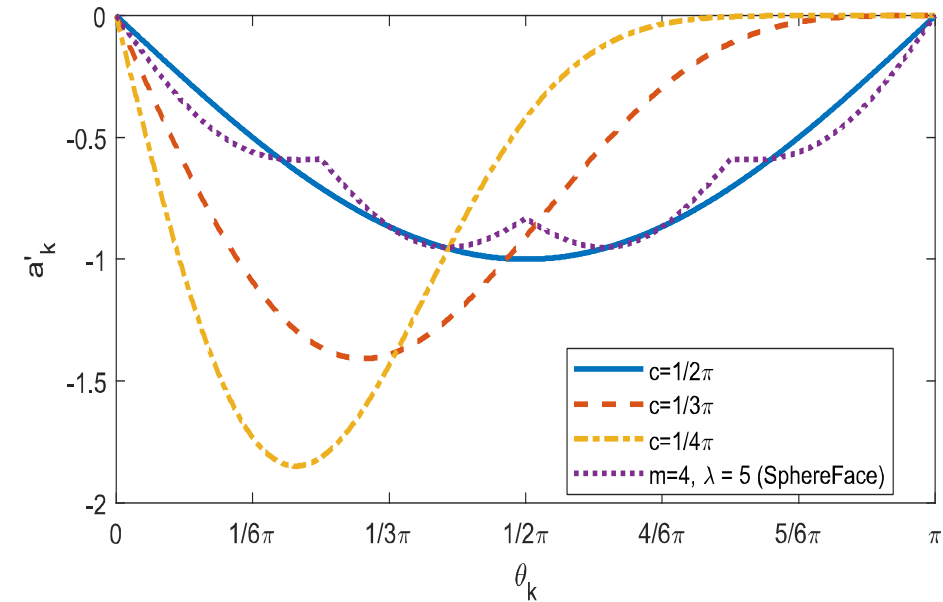
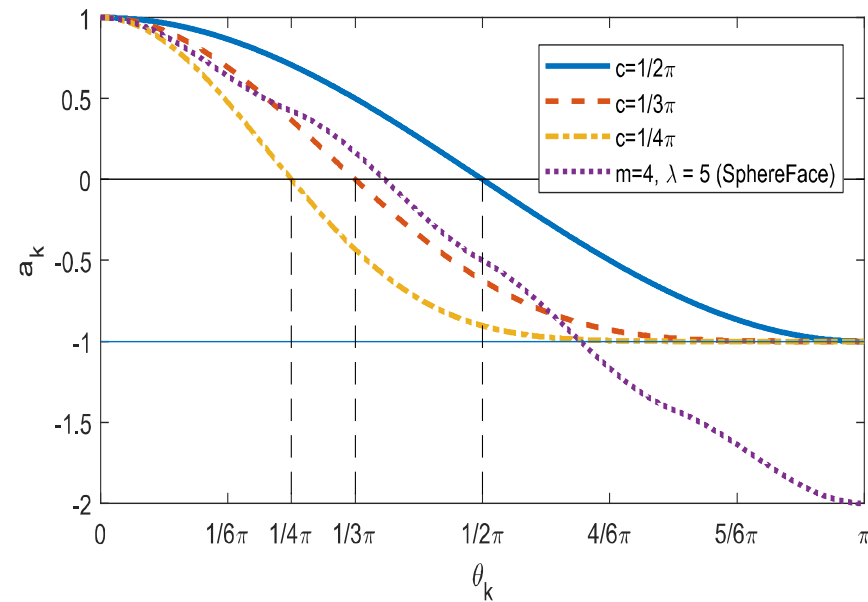


How well the model can recognize identities in the gallery from the probe set?

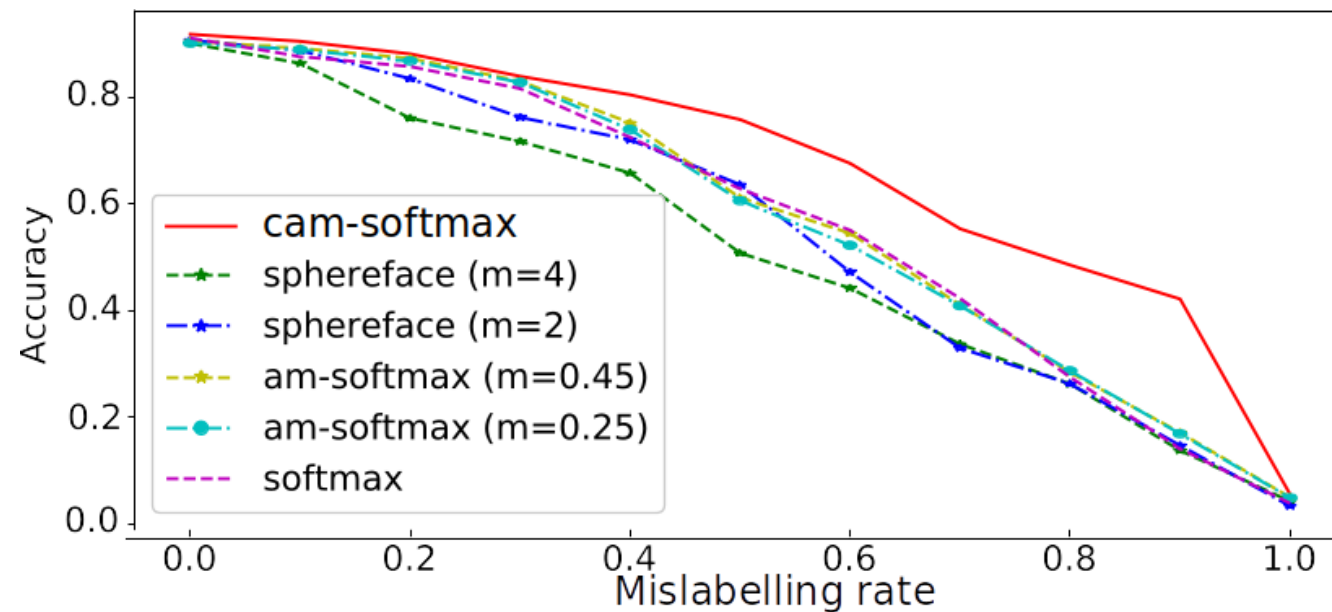




# Intuition for noisy labels

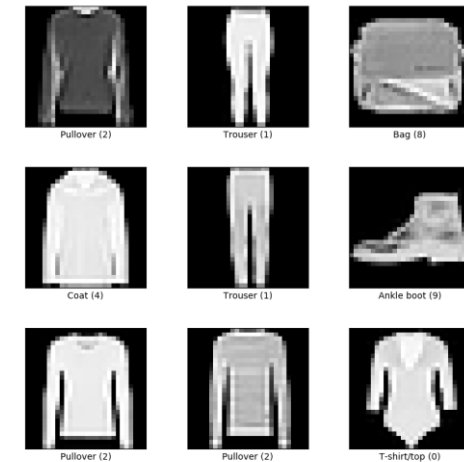


# Accuracy versus mislabelling rate on the Fashion-MNIST dataset



(The mislabelling is not uniformly random)

Fashion-MNIST examples



# Conclusion

- Better representation learning.
- Good for open-set recognition.
- Graceful degradation in the presence of label noise.