

INTRODUCTION

- Propose an approach to enhance the discriminative property of features for incremental classifier learning
- Build a network for the universal feature space in which a group of image classes have intra-class compactness and inter-class separability
- Model each incremental class to have a maximum margin from the rest of the models in universal space
- Experiments on CIFAR-100 dataset and IMDB-CAA demonstrate the superiority of our approach.

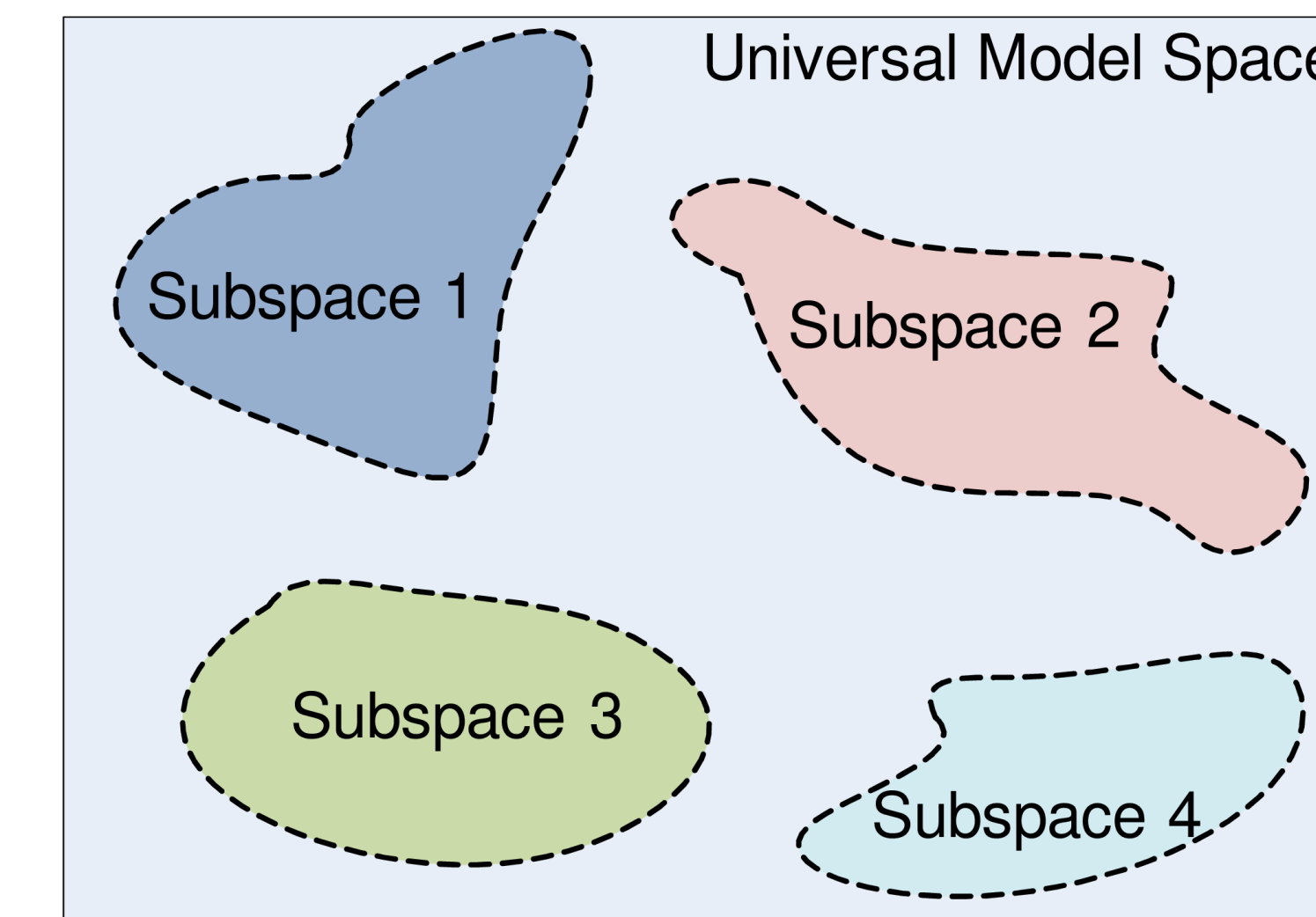
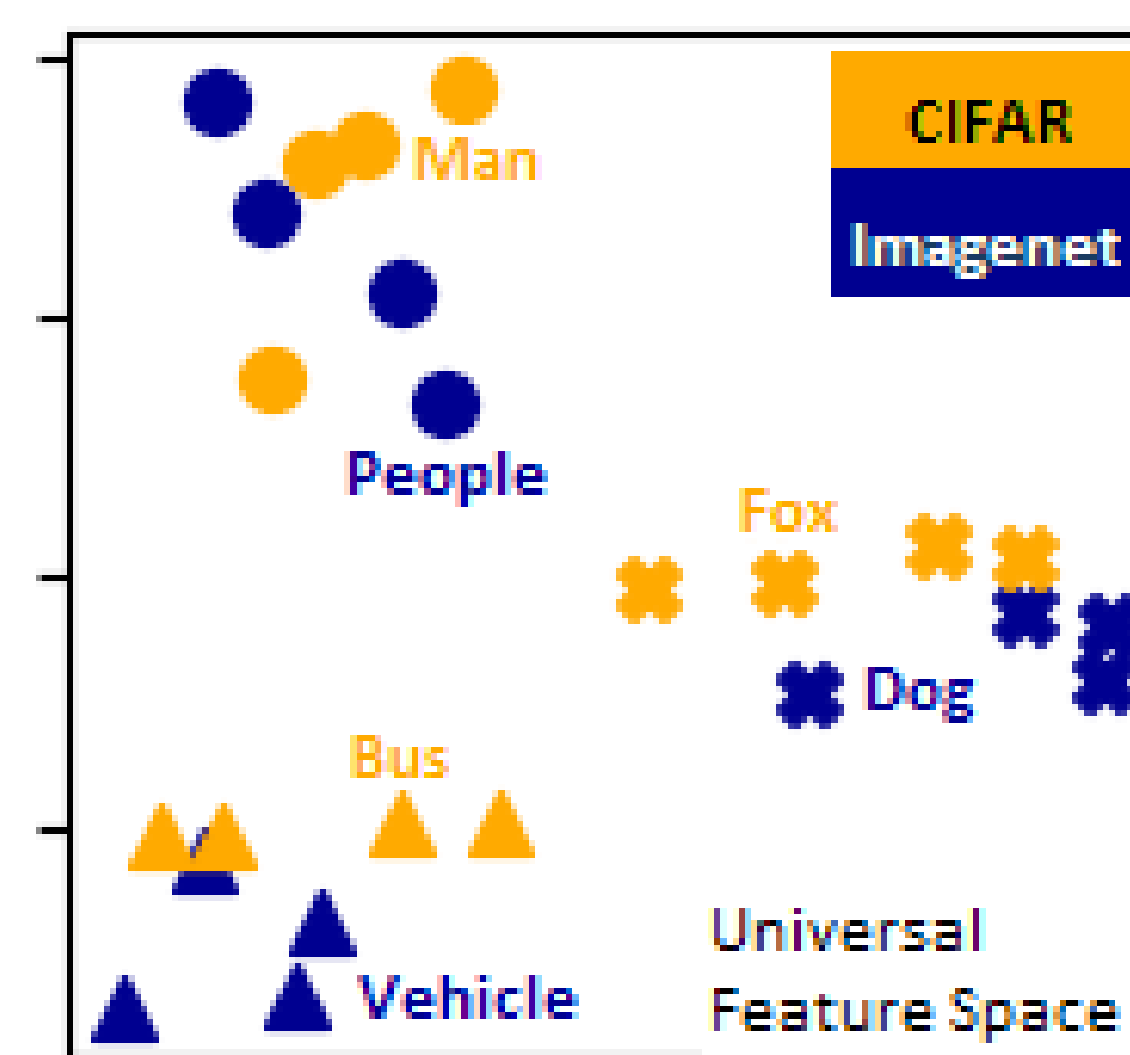
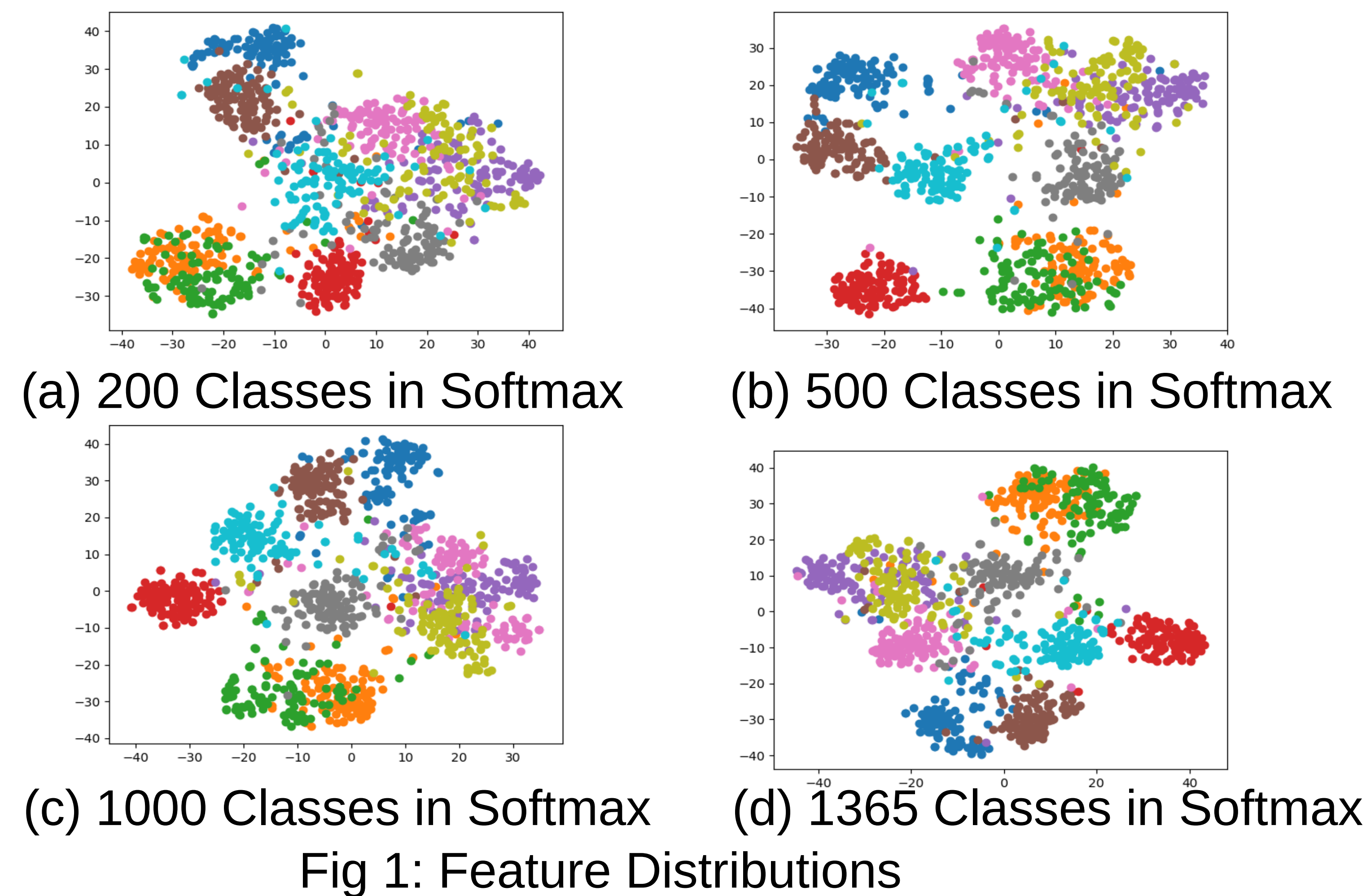
DISCRIMINATIVE FEATURES IN UFS

- ResNet-152 model trained using large Imagenet dataset as network for Universal Feature Space (UFS).
- Project classes of CIFAR-100 images into UFS.
- Include extra classes to the existing feature space of ResNet-152 for intra-class compactness and inter-class separability.
- ResNet-152 pre-trained model having 200 classes in Softmax layer to train a network for universal feature space.
- Extend the numbers of classes in Softmax layer to 500, 1000 and 1365 classes.

CLASSIFIER LEARNING IN UMS

- Create universal dataset and train binary SVM models for each class belonging to a subspace within a Universal Model Space (UMS)
- Optimize the margin between the region belonging to individual class and the rest in UMS

FEATURE AND MODEL DISTRIBUTIONS



DATASETS

- CIFAR-100 contains 100 image classes. Each class has 500 training images and 100 testing images.
- IMDB-CAA (Image Database for Context Aware Advertisement) has about 12,000 color or gray-scale images with all images are manually labeled.
- Each category has 500 training images and 100 testing images.

EXPERIMENTS

- Experiments on standard incremental learning and few-shot learning setup
- For each new class, obtain the discriminative features from the bottleneck layer of the UFS network.
- System learns each new class sequentially and we throw away all the samples after training binary SVM models.

Table 1: Standard Incremental Classifier Accuracies

Dataset	Numbers of classes in softmax layer			
	200	500	1000	1365
CIFAR-100	52	56	57	58.5
IMDB-CAA	53.9	61.2	71.7	72.9

Table 2: Few-shot Incremental Classifier Accuracies

Dataset	1000 Classes		1365 Classes	
	20-shots	15-shots	20-shots	15-shots
CIFAR-100	41.1	37.3	43.3	42
IMDB-CAA	65.2	64.4	72.6	65.9

CONCLUSION

- Proposed Incremental learning system has no re-training process and no storage requirement of old samples.
- System can accommodate incremental steps size as small as one to learning new classes.