Selected topics in Al 2 Assignment 2

| Name | ID | |
|---------------------------|----------|--|
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In this report we have constructed a classification model using active learning strategies applied on three datasets, two datasets are normally distributed while the third is unbalanced.

First, we'll give a brief regarding the datasets and the strategies used then we advance to the analysis and the information extracted from these scenarios.

Dataset brief

- 1. MNIST dataset: It consists of a collection of 70,000 grayscale images of handwritten digits, each of size 28 x 28 pixels. The dataset is split into two parts: 60,000 images for training and 10,000 images for testing.
- 2. Mushroom datasets: The dataset consists of 60,000 color images of size 32 x 32 pixels, divided into 10 classes with 6,000 images per class. The classes in the CIFAR dataset are: airplane, automobile, bird, cat, deer, dog, frog, horse, ship, and truck. Like the MNIST dataset, CIFAR is often used as a benchmark for evaluating the performance of machine learning algorithms on image classification tasks.
- 3. SVHN Dataset: consists of over 600,000 images, each of size 32 x 32 pixels, and is divided into three sets: a training set of 73257 images, a test set of 26032 images, and an extra set of 531131 images. The digits in the images can appear in various sizes, colors, fonts, and orientations, and can be located anywhere within the image. Next, we discuss the four query strategies that were used.

Query strategies

- Least confidence Sampling: This strategy selects the instances where the model is
 most uncertain about the predicted class label. The query strategy queries the
 instances for which the classifier is least confident. In the code, it is implemented by
 selecting the samples with the least confidence score as measured by the maximum
 probability of the predicted class. This strategy helps to maximize the model's
 accuracy by focusing on the most informative samples.
- 2. Entropy Sampling: This strategy measures the uncertainty of the predicted probability distribution of the classes. In other words, the query strategy chooses samples for which the model is most uncertain about the predicted probability distribution of the classes. In the code, it is implemented by selecting the samples with the highest entropy as measured by the predicted probability distribution. This strategy is useful when the model has low confidence or exhibits poor performance, but the samples it identifies are diverse and can help to better understand the data.
- 3. Random Sampling: This strategy selects the samples randomly from the unlabeled dataset, without any bias towards the informative samples. In the code, it is implemented by randomly selecting an instance from the pool of unlabeled samples. This strategy is useful when the dataset is uniformly distributed and there is no specific structure or pattern in the data.

- 4. Margin Sampling: This strategy selects the samples based on the difference in the probability of the predicted classes. The query strategy chooses samples for which the difference in the probability of the predicted classes is the smallest. In the code, it is implemented by selecting the samples with the smallest margin as measured by the predicted probability difference. This strategy is useful when the model has high accuracy and exhibits a high degree of confidence, but the samples it identifies are biased towards the most informative samples.
- 5. BALDdrop out sampling: "Bald" (acronym for "Bayesian active learning by disagreement") drop out sampling is a technique used in deep learning models to estimate model uncertainty. In deep learning, dropout is a regularization technique used to prevent overfitting. It involves randomly "dropping out" (i.e., setting to zero) a fraction of the neurons during training. This forces the remaining neurons to learn more robust features and reduces the reliance on any one neuron. "Bald" dropout extends this technique by using dropout during inference (i.e., when making predictions) and using the resulting variability in the predictions to estimate model uncertainty. Specifically, it involves computing the entropy (a measure of uncertainty) of the predicted probabilities over a set of dropout masks, and then selecting the samples with the highest entropy. The idea is that if the model is uncertain about a particular prediction, then it will give different results when different neurons are dropped out. By sampling from the dropout masks and selecting the samples with the highest entropy, we can estimate the uncertainty of the model's predictions. This technique has been shown to be effective in various applications, such as image classification, segmentation, and object detection, among others.

Now onto the analysis of the experiment,

MNIST Analysis

We picked 10000 random samples from dataset as initialized dataset then conducted a pool-based sampling approach for training and deep model for prediction. The model achieved 0.954 without active learning after applying it here are the results.

1. random sampling

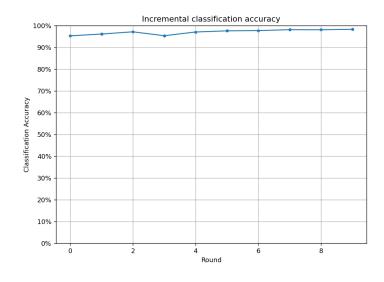
```
{'dataset_name': 'MNIST',
   'n_init_labeled': 10000,
   'n_query': 1000,
   'n_round': 10,
   'seed': 1,
   'strategy_name': 'RandomSampling'}
Train size: 60000 , Test size: 10000
number of labeled pool: 10000
number of unlabeled pool: 50000
number of testing pool: 10000
Round 0
```

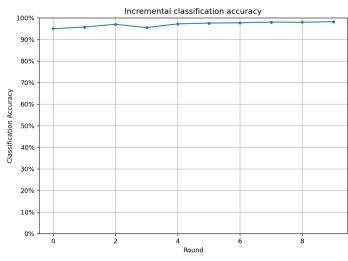
```
| 10/10 [00:33<00:00, 3.37s/it]
Round 0 trainig accuracy: 0.9541
Round 0 testing accuracy: 0.9562
Round 1
100%|
| 10/10 [00:36<00:00, 3.62s/it]
Round 1 training accuracy: 0.9532166666666667
Round 1 testing accuracy: 0.9564
Round 2
100%|
| 10/\overline{10} [00:40<00:00, 4.02s/it]
Round 2 training accuracy: 0.95895
Round 2 testing accuracy: 0.9601
Round 3
100%|
| 10/\overline{10} [00:43<00:00, 4.37s/it]
Round 3 training accuracy: 0.9584833333333334
Round 3 testing accuracy: 0.9617
Round 4
100%|
| 10/10 [00:47<00:00, 4.73s/it]
Round 4 training accuracy: 0.9615
Round 4 testing accuracy: 0.9668
Round 5
100%|
| 10/10 [00:50<00:00, 5.07s/it]
Round 5 training accuracy: 0.9646666666666667
Round 5 testing accuracy: 0.965
Round 6
100%|
| 10/10 [00:53<00:00, 5.35s/it]
Round 6 training accuracy: 0.9654333333333334
Round 6 testing accuracy: 0.9686
Round 7
100%|
                       5.74s/it]
| 10/10 [00:57<00:00,
Round 7 training accuracy: 0.9706166666666667
Round 7 testing accuracy: 0.9728
Round 8
100%|
10/10 [00:59<00:00, 5.94s/it]
Round 8 training accuracy: 0.96473333333333333
Round 8 testing accuracy: 0.9663
Round 9
1008|
| 10/10 [01:03<00:00, 6.35s/it]
Round 9 training accuracy: 0.9683333333333334
Round 9 testing accuracy: 0.9719
Round 10
100%|
| 10/\overline{10} [01:06<00:00, 6.68s/it]
Round 10 training accuracy: 0.9703333333333334
Round 10 testing accuracy: 0.9716
```

2. Entropy sampling

we used 25 queries instead of 30 as the first because it worsens after this number and this is the best strategy in terms of final acc

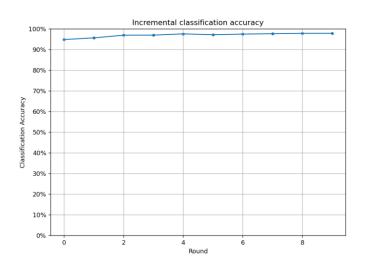
```
'dataset name': 'MNIST',
   'n init labeled': 5000,
   'n query': 1000,
   'n round': 10,
  'seed': 1,
  'strategy_name': 'EntropySampling'}
Train size: 60000 , Test size: 10000
number of labeled pool: 5000
number of unlabeled pool: 55000
number of testing pool: 10000
Round 0
100%|
| 10/10 [00:17<00:00, 1.80s/it]
Round 0 trainig accuracy: 0.9285833333333333
Round 0 testing accuracy: 0.934
Round 1
100%|
Round 1 training accuracy: 0.9487
Round 1 testing accuracy: 0.9523
Round 2
100%|
| 10/10 [00:23<00:00, 2.34s/it]
Round 2 training accuracy: 0.9565666666666667
Round 2 testing accuracy: 0.9607
Round 3
100%|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 2001|| | 20
Round 3 training accuracy: 0.9692166666666666
Round 3 testing accuracy: 0.9708
Round 4
100%|
| 10/10 [00:30<00:00, 3.05s/it]
Round 4 training accuracy: 0.9540166666666666
Round 4 testing accuracy: 0.953
Round 5
100%|
| 10/10 [00:32<00:00, 3.28s/it]
Round 5 training accuracy: 0.9712833333333334
Round 5 testing accuracy: 0.97
Round 6
100%|
| 10/10 [00:36<00:00, 3.65s/it]
Round 6 training accuracy: 0.9748166666666667
Round 6 testing accuracy: 0.9753
Round 7
100%|
| 10/10 [00:39<00:00, 3.94s/it]
Round 7 training accuracy: 0.9764333333333334
Round 7 testing accuracy: 0.9765
Round 8
```

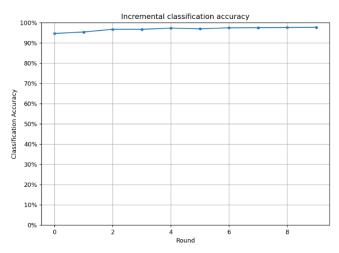




3. Bald sampling

```
'dataset name': 'MNIST',
 'n_init_labeled': 5000,
 'n query': 1000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'BALDDropout'}
Train size: 60000 , Test size: 10000
number of labeled pool: 5000
number of unlabeled pool: 55000
number of testing pool: 10000
Round 0
100%|
| 10/10 [00:17<00:00, 1.74s/it]
Round 0 trainig accuracy: 0.9285833333333333
Round 0 testing accuracy: 0.934
Round 1
100%|
| 10/10 [00:20<00:00, 2.06s/it]
Round 1 training accuracy: 0.9461
Round 1 testing accuracy: 0.9477
Round 2
100%|
Round 2 training accuracy: 0.9534
Round 2 testing accuracy: 0.9557
Round 3
10081
| 10/10 [00:26<00:00, 2.64s/it]
Round 3 training accuracy: 0.9668
Round 3 testing accuracy: 0.9687
Round 4
100%| 100:30<00:00, 3.01s/it]
Round 4 training accuracy: 0.96661666666666667
Round 4 testing accuracy: 0.969
Round 5
100%|
| 10/\overline{10} [00:33<00:00, 3.31s/it]
Round 5 training accuracy: 0.9727
Round 5 testing accuracy: 0.9751
Round 6
100%|
| 10/10 [00:36<00:00, 3.68s/it]
Round 6 training accuracy: 0.9690833333333333
Round 6 testing accuracy: 0.9709
Round 7
100%|
| 10/\overline{10} [00:39<00:00, 3.92s/it]
Round 7 training accuracy: 0.974
Round 7 testing accuracy: 0.974
Round 8
100%|
| 10/10 [00:43<00:00, 4.32s/it]
Round 8 training accuracy: 0.9748166666666667
```

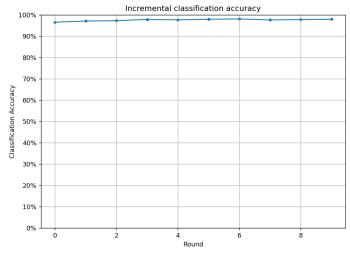


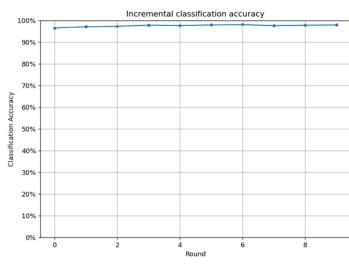


4. Least-confidence sampling

```
'dataset_name': 'MNIST',
'n_init_labeled': 5000,
 'n_query': 1000,
'n_round': 10,
 'seed': 1,
 'strategy name': 'LeastConfidence'}
Train size: 60000 , Test size: 10000
number of labeled pool: 5000
number of unlabeled pool: 55000
number of testing pool: 10000
Round 0
100%|
| 10/10 [00:18<00:00, 1.82s/it]
Round 0 testing accuracy: 0.934
Round 1
100%|
| 10/10 [00:20<00:00, 2.07s/it]
Round 1 training accuracy: 0.95063333333333333
Round 1 testing accuracy: 0.9539
Round 2
```

```
10/10 [00:23<00:00,
                       2.40s/it]
Round 2 training accuracy: 0.9601333333333333
Round 2 testing accuracy: 0.961
Round 3
100%|
| 10/10 [00:26<00:00, 2.69s/it]
Round 3 training accuracy: 0.96835
Round 3 testing accuracy: 0.9693
Round 4
100%|
10/10 [00:30<00:00,
                       3.06s/it]
Round 4 training accuracy: 0.9643166666666667
Round 4 testing accuracy: 0.9651
Round 5
100%|
| 10/10 [00:34<00:00,
                       3.43s/it]
Round 5 training accuracy: 0.97295
Round 5 testing accuracy: 0.9739
Round 6
100%|
10/10 [00:36<00:00,
Round 6 training accuracy: 0.9763
Round 6 testing accuracy: 0.9767
Round 7
100%|
| 10/10 [00:40<00:00, 4.09s/it]
Round 7 training accuracy: 0.9794833333333334
Round 7 testing accuracy: 0.9797
Round 8
100%|
| 10/10 [00:43<00:00,
                       4.36s/itl
Round 8 training accuracy: 0.9787833333333333
Round 8 testing accuracy: 0.9802
Round 9
100%|
10/10 [00:47<00:00,
                       4.72s/it]
Round 9 training accuracy: 0.97975
Round 9 testing accuracy: 0.9799
Round 10
100%|
| 10/\overline{10} | [00:52<00:00,
                       5.27s/it]
Round 10 training accuracy: 0.97973333333333333
Round 10 testing accuracy: 0.9811
```





CIFAR dataset

We picked 10000 random samples from dataset as initialized dataset) then conducted a pool-based sampling approach for training and Deep Classifier for prediction. The model achieved 0.594 without active learning after applying it here are the results.

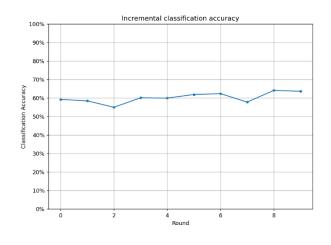
1. random sampling

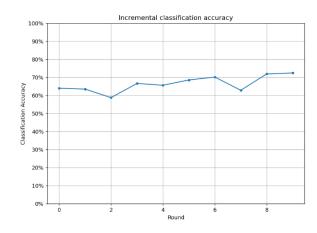
```
'dataset name': 'CIFAR10',
 'n init labeled': 10000,
 'n query': 1000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'RandomSampling'}
Files already downloaded and verified
Files already downloaded and verified
Train size: 50000 , Test size: 10000
number of labeled pool: 10000
number of unlabeled pool: 40000
number of testing pool: 10000
Round 0
100%|
 | 20/20 [01:22<00:00,
                          4.14s/it]
Round 0 trainig accuracy: 0.59434
Round 0 testing accuracy: 0.5499
Round 1
100%|
20/20 [01:29<00:00,
Round 1 training accuracy: 0.62666
Round 1 testing accuracy: 0.5898
Round 2
100%|
 | 20/20 [01:37<00:00,
                          4.86s/it]
Round 2 training accuracy: 0.63734
Round 2 testing accuracy: 0.5828
Round 3
100%| 20/20 [01:44<00:00, 5.24s/it]
Round 3 training accuracy: 0.44504
Round 3 testing accuracy: 0.4255
Round 4
100%|
20/20 [01:55<00:00,
                          5.77s/it]
Round 4 training accuracy: 0.65722
Round 4 testing accuracy: 0.6084
Round 5
100%|
  | 20/20 [02:03<00:00,</pre>
Round 5 training accuracy: 0.613
Round 5 testing accuracy: 0.5726
Round 6
100%|
   | 20/20 [02:09<00:00,
```

```
Round 6 training accuracy: 0.6869
Round 6 testing accuracy: 0.6275
Round 7
100%|
| 20/20 [02:16<00:00, 6.81s/it]
Round 7 training accuracy: 0.68188
Round 7 testing accuracy: 0.6221
Round 8
100%|
20/20 [02:25<00:00, 7.27s/it]
Round 8 training accuracy: 0.66668
Round 8 testing accuracy: 0.6077
Round 9
100%|
| 20/20 [02:33<00:00, 7.70s/it]
Round 9 training accuracy: 0.71812
Round 9 testing accuracy: 0.6455
Round 10
100%| 20/20 [02:50<00:00, 8.52s/it]
Round 10 training accuracy: 0.71924
Round 10 testing accuracy: 0.6538
  2. Entropy sampling
'dataset_name': 'CIFAR10',
'n_init_labeled': 10000,
 'n_query': 1000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'EntropySampling'}
Files already downloaded and verified
Files already downloaded and verified
Train size: 50000 , Test size: 10000
number of labeled pool: 10000
number of unlabeled pool: 400\overline{00}
number of testing pool: 10000
Round 0
100%|
20/20 [01:30<00:00, 4.50s/it]
Round 0 trainig accuracy: 0.59434
Round 0 testing accuracy: 0.5499
Round 1
100%|
| 20/20 [01:31<00:00, 4.56s/it]
Round 1 training accuracy: 0.63936
Round 1 testing accuracy: 0.5918
Round 2
100%| 20/20 [01:38<00:00, 4.93s/it]
```

Round 2 training accuracy: 0.63484

```
Round 2 testing accuracy: 0.5845
Round 3
100%|
 20/20 [01:48<00:00,
                         5.42s/itl
Round 3 training accuracy: 0.58748
Round 3 testing accuracy: 0.5495
Round 4
100%|
20/20 [01:54<00:00, 5.73s/it]
Round 4 training accuracy: 0.66602
Round 4 testing accuracy: 0.6007
Round 5
100%|
20/20 [02:04<00:00, 6.24s/it]
Round 5 training accuracy: 0.65604
Round 5 testing accuracy: 0.599
Round 6
100%|| 20/20 [02:12<00:00, 6.65s/it]
Round 6 training accuracy: 0.6853
Round 6 testing accuracy: 0.6186
Round 7
100%|
| 20/20 [02:19<00:00,
                         6.97s/itl
Round 7 training accuracy: 0.70128
Round 7 testing accuracy: 0.6233
Round 8
100%|
 | 20/20 [02:27<00:00,
                         7.38s/it]
Round 8 training accuracy: 0.6283
Round 8 testing accuracy: 0.5775
Round 9
100%|
20/20 [02:37<00:00, 7.86s/it]
Round 9 training accuracy: 0.71886
Round 9 testing accuracy: 0.6409
Round 10
100%|
| 20/20 [02:43<00:00, 8.17s/it]
Round 10 training accuracy: 0.72396
Round 10 testing accuracy: 0.6361
```

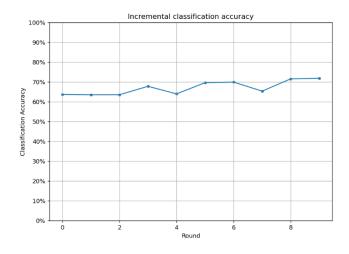


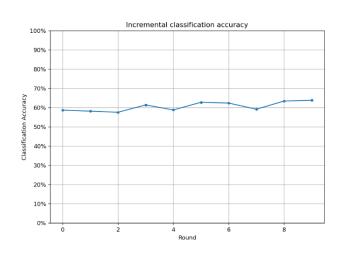


3. BALD sampling

```
'dataset_name': 'CIFAR10',
'n_init_labeled': 10000,
 'n_query': 1000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'BALDDropout'}
Files already downloaded and verified
Files already downloaded and verified
Train size: 50000 , Test size: 10000
number of labeled pool: 10000
number of unlabeled pool: 40000
number of testing pool: 10000
Round 0
100%|
20/20 [01:22<00:00,
                         4.12s/it]
Round 0 trainig accuracy: 0.59434
Round 0 testing accuracy: 0.5499
Round 1
100%|
20/20 [01:30<00:00, 4.53s/it]
Round 1 training accuracy: 0.63664
Round 1 testing accuracy: 0.5862
Round 2
100%| 20/20 [01:37<00:00, 4.86s/it]
Round 2 training accuracy: 0.63496
Round 2 testing accuracy: 0.581
Round 3
100%
| 20/20 [01:44<00:00, 5.24s/it]
Round 3 training accuracy: 0.63522
Round 3 testing accuracy: 0.5753
Round 4
100%|
| 20/20 [01:54<00:00, 5.71s/it]
Round 4 training accuracy: 0.67758
Round 4 testing accuracy: 0.6134
Round 5
100%|
20/20 [02:05<00:00, 6.27s/it]
Round 5 training accuracy: 0.63958
Round 5 testing accuracy: 0.5874
Round 6
100%|
| 20/20 [02:13<00:00, 6.66s/it]
```

```
Round 6 training accuracy: 0.69592
Round 6 testing accuracy: 0.6272
Round 7
100%|
| 20/20 [02:21<00:00,
Round 7 training accuracy: 0.69882
Round 7 testing accuracy: 0.623
Round 8
100%| 20/20 [02:31<00:00, 7.56s/it]
Round 8 training accuracy: 0.65328
Round 8 testing accuracy: 0.5911
Round 9
100%|
| 20/20 [02:36<00:00,
                          7.85s/it]
Round 9 training accuracy: 0.71526
Round 9 testing accuracy: 0.6336
Round 10
100%| 20/20 [02:46<00:00,
                          8.30s/it]
Round 10 training accuracy: 0.7182
Round 10 testing accuracy: 0.6375
```



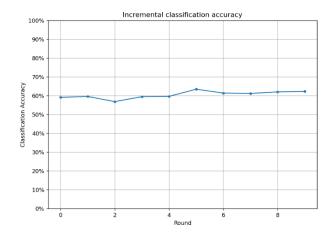


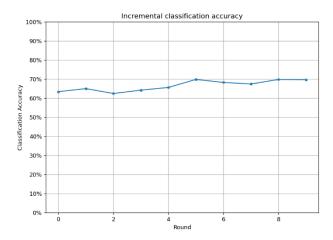
4. least con. sampling

```
{'dataset_name': 'CIFAR10',
  'n_init_labeled': 10000,
  'n_query': 1000,
  'n_round': 10,
  'seed': 1,
  'strategy_name': 'LeastConfidence'}
```

```
Files already downloaded and verified Files already downloaded and verified Train size: 50000, Test size: 10000 number of labeled pool: 10000 number of unlabeled pool: 40000 number of testing pool: 10000
```

```
Round 0
100%|
 20/20 [01:24<00:00,
                         4.25s/itl
Round 0 trainig accuracy: 0.59434
Round 0 testing accuracy: 0.5499
Round 1
100%|
20/20 [01:33<00:00, 4.67s/it]
Round 1 training accuracy: 0.63382
Round 1 testing accuracy: 0.5909
Round 2
100%|
20/20 [01:39<00:00, 4.98s/it]
Round 2 training accuracy: 0.64988
Round 2 testing accuracy: 0.5958
Round 3
100%|| 20/20 [01:48<00:00, 5.41s/it]
Round 3 training accuracy: 0.62428
Round 3 testing accuracy: 0.5683
Round 4
100%|
| 20/20 [01:55<00:00, 5.77s/it]
Round 4 training accuracy: 0.64186
Round 4 testing accuracy: 0.5945
Round 5
100%|
20/20 [02:05<00:00,
                         6.26s/it]
Round 5 training accuracy: 0.65604
Round 5 testing accuracy: 0.5963
Round 6
100%|
20/20 [02:13<00:00, 6.68s/it]
Round 6 training accuracy: 0.69796
Round 6 testing accuracy: 0.6341
Round 7
100%|
| 20/20 [02:18<00:00, 6.92s/it]
Round 7 training accuracy: 0.6823
Round 7 testing accuracy: 0.6138
Round 8
100%|
20/20 [02:26<00:00, 7.34s/it]
Round 8 training accuracy: 0.67398
Round 8 testing accuracy: 0.6117
Round 9
100%|
| 20/20 [02:36<00:00, 7.81s/it]
Round 9 training accuracy: 0.6977
Round 9 testing accuracy: 0.62
Round 10
100%| 20/20 [02:46<00:00, 8.31s/it]
Round 10 training accuracy: 0.69668
Round 10 testing accuracy: 0.6223
```





5. margin sampling

```
{'dataset_name': 'CIFAR10',
  'n_init_labeled': 10000,
  'n_query': 1000,
  'n_round': 10,
  'seed': 1,
  'strategy_name': 'MarginSampling'}
```

Files already downloaded and verified Files already downloaded and verified Train size: 50000, Test size: 10000 number of labeled pool: 10000 number of unlabeled pool: 40000 number of testing pool: 10000

Round 4 testing accuracy: 0.6164

```
Round 0
100%|
| 20/20 [01:21<00:00, 4.08s/it]
Round 0 trainig accuracy: 0.59434
Round 0 testing accuracy: 0.5499
Round 1
100%|
20/20 [01:30<00:00,
                         4.51s/it]
Round 1 training accuracy: 0.63318
Round 1 testing accuracy: 0.5899
Round 2
100%| 20/20 [01:38<00:00, 4.92s/it]
Round 2 training accuracy: 0.63244
Round 2 testing accuracy: 0.5892
Round 3
100%|
| 20/20 [01:46<00:00, 5.32s/it]
Round 3 training accuracy: 0.64122
Round 3 testing accuracy: 0.5941
Round 4
100%|
20/20 [01:55<00:00,
                         5.76s/it]
Round 4 training accuracy: 0.67854
```

```
Round 5
100%|
  | 20/20 [02:01<00:00,
                          6.08s/it]
Round 5 training accuracy: 0.66368
Round 5 testing accuracy: 0.6081
Round 6
100%|
  | 20/20 [02:09<00:00,
                          6.48s/it]
Round 6 training accuracy: 0.69538
Round 6 testing accuracy: 0.6286
Round 7
100%|
  | 20/20 [02:17<00:00,
Round 7 training accuracy: 0.67266
Round 7 testing accuracy: 0.6136
Round 8
100%| 20/20 [02:26<00:00,
Round 8 training accuracy: 0.68638
Round 8 testing accuracy: 0.6205
Round 9
100%|
 20/20 [02:33<00:00,
Round 9 training accuracy: 0.71036
Round 9 testing accuracy: 0.6346
Round 10
100%| 20/20 [02:41<00:00,
                          8.07s/it]
Round 10 training accuracy: 0.69514
Round 10 testing accuracy: 0.6173
```

SVHN dataset



The SVHN dataset is not perfectly balanced, but it is fairly balanced compared to some other datasets. In particular, each digit class has roughly the same number of examples in the training set, with a slight skew towards the number 1, which has slightly more examples than the other digits. The test set is also fairly balanced, with a similar distribution of digits as the training set.

The extra set of images in the SVHN dataset is less balanced than the training and test sets and contains a larger proportion of images with the number 1, as well as some other biases.

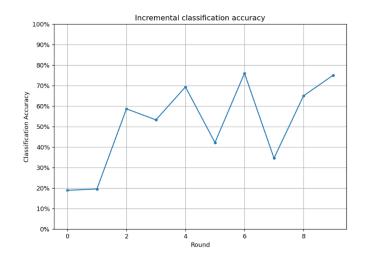
This is partly because the extra set was created by applying weakly supervised labeling to a large number of unlabeled images, which can introduce biases in the labeling process.

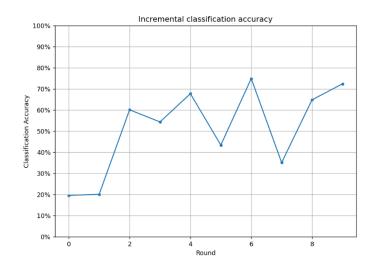
Overall, the SVHN dataset is not perfectly balanced and some consider unbalanced

1 Least-confidence sampling:

```
'dataset name': 'SVHN',
 'n init labeled': 5000,
 'n query': 1000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'LeastConfidence'}
Using downloaded and verified file: ./data/SVHN/train 32x32.mat
Using downloaded and verified file: ./data/SVHN/test 32x32.mat
Train size: 73257 , Test size: 26032
number of labeled pool: 5000
number of unlabeled pool: 65000
number of testing pool: 26032
Round 0
100%|
                       2.27s/it]
Round 0 trainig accuracy: 0.18958571428571427
Round 0 testing accuracy: 0.19598955132145052
Round 1
100%|
| 20/\overline{20} | [00:52<00:00,
                       2.62s/it]
Round 1 training accuracy: 0.1895428571428<u>5</u>715
Round 1 testing accuracy: 0.19591272280270436
Round 2
100%|
| 20/20 [01:01<00:00,
                       3.09s/itl
Round 2 training accuracy: 0.19587142857142859
Round 2 testing accuracy: 0.20090657652120467
Round 3
100%|
                       3.65s/it]
Round 3 training accuracy: 0.5865571428571429
Round 3 testing accuracy: 0.6008758451137062
Round 4
100%|
| 20/\overline{20} | [01:19<00:00,
                       4.00s/it]
Round 4 training accuracy: 0.5326857142857143
Round 4 testing accuracy: 0.5434465273509527
Round 5
100%|
                        4.42s/it]
Round 5 training accuracy: 0.6934571428571429
Round 5 testing accuracy: 0.6772049784880148
Round 6
100%|
| 20/20 [01:36<00:00, 4.81s/it]
Round 6 training accuracy: 0.4221714285714286
Round 6 testing accuracy: 0.4340043023970498
```

```
Round 7
100%|
                       5.27s/it]
20/20 [01:45<00:00,
Round 7 training accuracy: 0.7598428571428572
Round 7 testing accuracy: 0.7486555009219422
Round 8
100%|
| 20/\overline{20} | 01:53<00:00,
                       5.66s/it]
Round 8 training accuracy: 0.34587142857142855
Round 8 testing accuracy: 0.35122157344806393
Round 9
100%|
Round 9 training accuracy: 0.6498571428571429
Round 9 testing accuracy: 0.6481637984019668
Round 10
100%|
| 20/20 [02:09<00:00, 6.49s/it]
Round 10 training accuracy: 0.749942857142<u>8571</u>
Round 10 testing accuracy: 0.7236094038106945
```

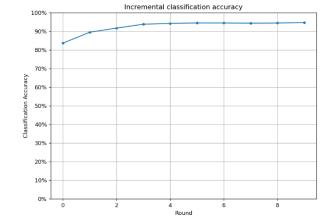


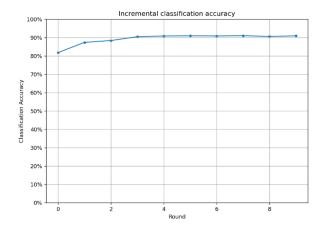


2 Bald sampling

```
'dataset name': 'SVHN',
 'n_init_labeled': 5000,
 'n_query': 10000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'BALDDropout'}
Train size: 73257 , Test size: 26032
number of labeled pool: 5000
number of unlabeled pool: 65000
number of testing pool: 26032
Round 0
100%|
| 20/20 [00:48<00:00,
Round 0 trainig accuracy: 0.18958571428571427
Round 0 testing accuracy: 0.19598955132145052
Round 1
```

```
100%|
 20/20 [02:13<00:00,
                       6.66s/it]
Round 1 training accuracy: 0.8356285714285714
Round 1 testing accuracy: 0.8176475107559926
Round 2
100%|
| 20/20 [03:41<00:00, 11.07s/it]
Round 2 training accuracy: 0.8943714285714286
Round 2 testing accuracy: 0.8733866011063307
Round 3
100%|
| 20/20 [05:06<00:00, 15.30s/it]
Round 3 training accuracy: 0.9164571428571429
Round 3 testing accuracy: 0.8834895513214506
Round 4
100%|
| 20/\overline{20} [06:35<00:00, 19.77s/it]
Round 4 training accuracy: 0.9373285714285714
Round 4 testing accuracy: 0.9041180086047941
Round 5
100%|
| 20/20 [07:59<00:00, 24.00s/it]
Round 5 training accuracy: 0.9416857142857142
Round 5 testing accuracy: 0.9080746773202213
Round 6
100%|
| 20/20 [09:23<00:00, 28.16s/it]
Round 6 training accuracy: 0.9441428571428572
Round 6 testing accuracy: 0.9092655193607867
Round 7
100%|
| 20/20 [10:04<00:00, 30.23s/it]
Round 7 training accuracy: 0.9440142857142857
Round 7 testing accuracy: 0.9082667486170867
Round 8
100%|
| 20/20 [10:04<00:00, 30.23s/it]
Round 8 training accuracy: 0.9428142857142857
Round 8 testing accuracy: 0.9100722188076213
Round 9
100%|
| 20/20 [10:05<00:00, 30.29s/it]
Round 9 training accuracy: 0.9438
Round 9 testing accuracy: 0.9053856791641057
Round 10
100%|
| 20/20 [10:02<00:00, 30.10s/it]
Round 10 training accuracy: 0.9463142857142857
Round 10 testing accuracy: 0.9088813767670559
```

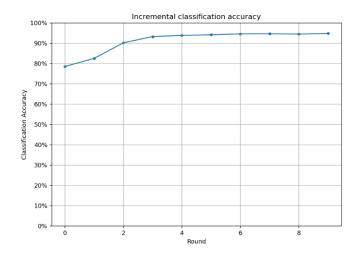


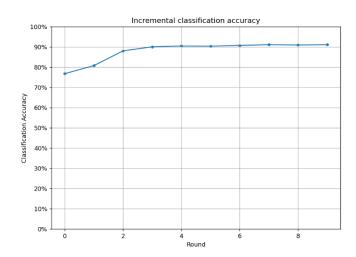


3 entropy sampling

```
'dataset_name': 'SVHN',
'n_init_labeled': 5000,
 'n_query': 10000,
 'n round': 10,
 'seed': 1,
 'strategy name': 'EntropySampling'}
Using downloaded and verified file: ./data/SVHN/train 32x32.mat
Using downloaded and verified file: ./data/SVHN/test\_32x32.mat
Train size: 73257 , Test size: 26032
number of labeled pool: 5000
number of unlabeled pool: 65000
number of testing pool: 26032
Round 0
100%|
| 20/\overline{20} [00:47<00:00, 2.37s/it]
Round 0 trainig accuracy: 0.18958571428571427
Round 0 testing accuracy: 0.19598955132145052
Round 1
100%|
| 20/20 [02:13<00:00, 6.70s/it]
Round 1 training accuracy: 0.7846
Round 1 testing accuracy: 0.7672864167178857
Round 2
100%
| 20/20 [03:41<00:00, 11.07s/it]
Round 2 training accuracy: 0.8242428571428572
Round 2 testing accuracy: 0.8070835894283959
Round 3
100%|
| 20/20 [05:10<00:00, 15.52s/it]
Round 3 training accuracy: 0.9008
Round 3 testing accuracy: 0.8798786109403811
Round 4
100%|
| 20/20 [06:39<00:00, 19.99s/it]
Round 4 training accuracy: 0.9319
Round 4 testing accuracy: 0.8999692685925015
Round 5
100%|
| 20/20 [08:08<00:00, 24.42s/it]
Round 5 training accuracy: 0.9372142857142857
Round 5 testing accuracy: 0.9039643515673018
Round 6
```

```
20/20 [09:30<00:00, 28.50s/it]
Round 6 training accuracy: 0.9405428571428571
Round 6 testing accuracy: 0.9032344806392133
Round 7
100%|
 20/20 [10:08<00:00, 30.41s/it]
Round 7 training accuracy: 0.9450142857142857
Round 7 testing accuracy: 0.9068070067609096
Round 8
100%|
20/20 [10:08<00:00, 30.41s/it]
Round 8 training accuracy: 0.9454
Round 8 testing accuracy: 0.9105716041794
Round 9
100%|
 20/20 [10:03<00:00, 30.19s/it]
Round 9 training accuracy: 0.943985714285714
Round 9 testing accuracy: 0.9089966195451752
Round 10
100%|
 20/20 [10:04<00:00, 30.23s/it]
Round 10 training accuracy: 0.9470285714285714
Round 10 testing accuracy: 0.9104563614013522
```





Conclusion

Each strategy in the uncertainty family doesn't vary much as we can't for sure say that one strategy is better than other overall however the active learning approaches as a whole are better in the factors of maintaining high score in evaluation metrics.

MNIST had the highest acc with low computing power while SVHN scored the highest variance in learning due to instability.