```
import scipy.io
In [1]:
        import pandas as pd
        import numpy as np
        import time
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.linear_model import LogisticRegression
        from sklearn.svm import SVC
        from sklearn.neural network import MLPClassifier
        from sklearn.metrics import accuracy_score, confusion_matrix, r
In [2]: data = scipy.io.loadmat('mnist_10digits.mat')
        #standardize data to [0,1]
        xtrain = data['xtrain']/255
        ytrain = data['ytrain'].ravel()
        xtest = data['xtest']/255
        ytest = data['ytest'].ravel()
In [3]:
        #set seed for downsampling
        seed = 55
        np.random.seed(seed)
In [4]:
        #downsample for knn and svm
        m = 5000
        ind = np.random.choice(xtrain.shape[0], m, replace=False)
        print(ind)
        xtrain_2 = xtrain[ind]
        ytrain_2 = ytrain[ind]
        [32593 59700 4764 ... 29518 39147 50296]
```

### KNN

```
In [5]: #find best k using accuracy as metric
start = time.time()

k_accs = {}

for i in range(1,6):
    knn_clf = KNeighborsClassifier(n_neighbors=i).fit(xtrain_2,
    knn_pred = knn_clf.predict(xtest)

# Find acc at each k
```

```
knn acc = accuracy score(ytest, knn pred)
            print("KNN Classifier Accuracy:", knn_acc, 'at k =', i)
            k accs[i] = knn acc
        end = time.time()
        print('Run time:', end-start)
        KNN Classifier Accuracy: 0.9386 at k = 1
        KNN Classifier Accuracy: 0.9246 at k = 2
        KNN Classifier Accuracy: 0.9371 at k = 3
        KNN Classifier Accuracy: 0.9345 at k = 4
        KNN Classifier Accuracy: 0.9377 at k = 5
        Run time: 7.14811110496521
        \max k = \max(k \text{ accs, key=k accs.get})
In [6]:
        print(max k)
In [7]:
        # Rerun knn using best k value
        knn clf = KNeighborsClassifier(n neighbors=max k).fit(xtrain 2,
        knn pred = knn clf.predict(xtest)
        #find metrics
        knn acc = accuracy score(ytest, knn pred)
        knn_cm = confusion_matrix(ytest, knn_pred)
        knn_precision = precision_score(ytest, knn_pred, average=None)
        knn_recall = recall_score(ytest, knn_pred, average=None)
        knn_f1 = f1_score(ytest, knn_pred, average=None)
```

# **Logistic Regression**

```
In [8]: #log reg model
    start = time.time()

    lr_clf = LogisticRegression(max_iter=10000).fit(xtrain, ytrain)

    lr_pred = lr_clf.predict(xtest)

    end = time.time()
    print('Run time:', end-start)

    Run time: 149.2913339138031

In [9]: #find metrics
    lr_acc = accuracy_score(ytest, lr_pred)
    lr_cm = confusion_matrix(ytest, lr_pred)
    lr_precision = precision_score(ytest, lr_pred, average=None)
```

```
lr_recall = recall_score(ytest, lr_pred, average=None)
lr_f1 = f1_score(ytest, lr_pred, average=None)
```

### **SVM**

```
#svm model
In [10]:
         start = time.time()
         svm_clf = SVC(kernel='linear').fit(xtrain_2, ytrain_2)
         svm_pred = svm_clf.predict(xtest)
         end = time.time()
         print('Run time:', end-start)
         Run time: 7.694239854812622
         #find metrics
In [11]:
         svm acc = accuracy_score(ytest, svm pred)
         svm cm = confusion matrix(ytest, svm pred)
         svm_precision = precision_score(ytest, svm_pred, average=None)
         svm_recall = recall_score(ytest, svm_pred, average=None)
         svm_f1 = f1_score(ytest, svm_pred, average=None)
         Kernal SVM
In [12]: #kernal svm model
         start = time.time()
         k_svm_clf = SVC(kernel='rbf').fit(xtrain_2, ytrain_2)
         k svm pred = k svm clf.predict(xtest)
         end = time.time()
         print('Run time:', end-start)
         Run time: 15.709005117416382
```

k\_svm\_acc = accuracy\_score(ytest, k\_svm\_pred)
k\_svm\_cm = confusion\_matrix(ytest, k\_svm\_pred)

k svm f1 = f1\_score(ytest, k\_svm\_pred, average=None)

k\_svm\_precision = precision\_score(ytest, k\_svm\_pred, average=Note)
k\_svm\_recall = recall\_score(ytest, k\_svm\_pred, average=None)

### **Neural Network**

#find metrics

In [13]:

```
In [14]:
         #Neural network model
         start = time.time()
         nn clf = MLPClassifier(hidden layer sizes=(20, 10)).fit(xtrain,
         nn_pred = nn_clf.predict(xtest)
         end = time.time()
         print('Run time:', end-start)
         Run time: 294,62807035446167
         /Users/yuxi/anaconda3/lib/python3.10/site-packages/sklearn/neu
         ral_network/_multilayer_perceptron.py:684: ConvergenceWarning:
         Stochastic Optimizer: Maximum iterations (200) reached and the
         optimization hasn't converged yet.
           warnings.warn(
In [15]: #find metrics
         nn_acc = accuracy_score(ytest, nn_pred)
         nn_cm = confusion_matrix(ytest, nn_pred)
         nn_precision = precision_score(ytest, nn_pred, average=None)
         nn_recall = recall score(ytest, nn_pred, average=None)
         nn_f1 = f1_score(ytest, nn_pred, average=None)
```

### Visualize Results

```
In [22]: def results(clf, cm, precision, recall, f1, accuracy):
    # show confusion matrix
    plt.figure(figsize=(5, 3))
    sns.heatmap(cm, annot=True, fmt='d', cmap='Reds', cbar=Fals
    plt.xlabel('Predicted Labels')
    plt.ylabel('True Labels')
    plt.title(f'{clf} Confusion Matrix')
    plt.savefig(f'{clf} Confusion Matrix')
    plt.show()

#show metrics
    df = pd.DataFrame({"Precision": precision, "Recall": recall, display(df)
In [23]: results("KNN", knn cm, knn precision, knn recall, knn f1, knn cm, knn precision, knn
```

### KNN Confusion Matrix True Labels

Predicted Labels

0.9386

0.9386

0.9386

**Precision** Recall F1-Score Accuracy 0.956219 0.980612 0.968262 0.9386 0.931631 0.996476 0.962963 0.9386 0.975916 0.903101 0.938098 0.9386 0.908566 0.934653 0.921425 0.9386 0.941176 0.9386 0.953975 0.928717 0.920354 0.932735 0.926503 0.9386 0.964803 0.968815 0.9386 0.972860

0.942607

0.869610

0.917740

0.930836

0.962500

0.890385

In [24]: results("Logistic Regression", lr\_cm, lr\_precision, lr\_recall,

0.936684

0.913700

0.903856

# Logistic Regression Confusion Matrix

True Labels	0	- 95	5	0	2	4	1	10	4	3	1	0
	J	- 0	)	1110	5	2	0	2	3	2	11	0
	7	- 6	ò	9	930	14	10	3	12	10	34	4
	Э	- 4	ļ	1	16	925	1	23	2	10	19	9
	4	- 1	L	3	7	3	921	0	6	5	6	30
	2	- 9	)	2	3	35	10	777	15	6	31	4
	9	- 8	3	3	8	2	6	16	912	2	1	0
	7	- 1	L	7	23	7	6	1	0	947	4	32
	œ	- 9	)	11	6	22	7	29	13	10	855	12
	6	- 9	)	8	1	9	21	7	0	21	9	924
		Ċ	)	'n	2	3	4	5	6	7	8	9
Predicted Labels												

	Precision	Recall	F1-Score	Accuracy
0	0.953094	0.974490	0.963673	0.9256
1	0.961872	0.977974	0.969856	0.9256
2	0.929071	0.901163	0.914904	0.9256
3	0.904203	0.915842	0.909985	0.9256
4	0.936928	0.937882	0.937405	0.9256
5	0.895161	0.871076	0.882955	0.9256
6	0.943123	0.951983	0.947532	0.9256
7	0.932087	0.921206	0.926614	0.9256
8	0.880536	0.877823	0.879177	0.9256
9	0.910345	0.915758	0.913043	0.9256

In [25]: results("SVM", svm\_cm, svm\_precision, svm\_recall, svm\_f1, svm\_a

## **SVM Confusion Matrix**

	0 -	956	0	4	3	0	7	7	1	1	1
	н -	0	1116	4	3	0	1	4	2	5	0
	7 -	9	7	935	22	17	2	13	15	8	4
5	m -	3	4	22	902	1	24	0	13	29	12
True Labels	4 -	1	4	5	3	920	0	9	4	3	33
	٦ -	18	4	14	54	9	747	17	3	20	6
	9 -	11	3	11	0	18	19	891	1	4	0
		1	12	33	14	16	0	0	923	0	29
	ω -	4	18	17	59	11	38	17	6	797	7
	ი -	5	11	4	11	53	7	1	29	10	878
		ó	'n	2	3	4	5	6	7	8	9
Predicted Labels											

	Precision	Recall	F1-Score	Accuracy
0	0.948413	0.975510	0.961771	0.9065
1	0.946565	0.983260	0.964564	0.9065
2	0.891325	0.906008	0.898606	0.9065
3	0.842204	0.893069	0.866891	0.9065
4	0.880383	0.936864	0.907745	0.9065
5	0.884024	0.837444	0.860104	0.9065
6	0.929093	0.930063	0.929577	0.9065
7	0.925777	0.897860	0.911605	0.9065
8	0.908780	0.818275	0.861156	0.9065
9	0.905155	0.870168	0.887317	0.9065

In [26]: results("Kernel SVM", k\_svm\_cm, k\_svm\_precision, k\_svm\_recall,

### Kernel SVM Confusion Matrix o - 965 True Labels

Predicted Labels

	Precision	Recall	F1-Score	Accuracy
0	0.966934	0.984694	0.975733	0.954
1	0.976419	0.985022	0.980702	0.954
2	0.951220	0.944767	0.947982	0.954
3	0.934109	0.954455	0.944172	0.954
4	0.939364	0.962322	0.950704	0.954
5	0.951192	0.939462	0.945290	0.954
6	0.954872	0.971816	0.963270	0.954
7	0.962889	0.933852	0.948148	0.954
8	0.948798	0.932238	0.940445	0.954
9	0.951220	0.927651	0.939288	0.954

In [27]: results("Neural Networks", nn\_cm, nn\_precision, nn\_recall, nn\_1

### Neural Networks Confusion Matrix o - 956 True Labels 930 0

Predicted Labels

	Precision	Recall	F1-Score	Accuracy
0	0.959839	0.975510	0.967611	0.9488
1	0.971354	0.985903	0.978575	0.9488
2	0.938154	0.955426	0.946711	0.9488
3	0.945545	0.945545	0.945545	0.9488
4	0.951894	0.947047	0.949464	0.9488
5	0.938286	0.920404	0.929259	0.9488
6	0.945248	0.955115	0.950156	0.9488
7	0.959526	0.945525	0.952474	0.9488
8	0.940439	0.924025	0.932160	0.9488
9	0.933067	0.925669	0.929353	0.9488