Exercise 9: Support Vector Machines

CPSC 381/581: Machine Learning

Yale University

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Prerequisites:

- 1. Enable Google Colaboratory as an app on your Google Drive account
- 2. Create a new Google Colab notebook, this will also create a "Colab Notebooks" directory under "MyDrive" i.e.

/content/drive/MyDrive/Colab Notebooks

3. Create the following directory structure in your Google Drive

/content/drive/MyDrive/Colab Notebooks/CPSC 381-581: Machine Learning/Exercises

4. Move the 09_exercise_support_vector_machines.ipynb into

/content/drive/MyDrive/Colab Notebooks/CPSC 381-581: Machine Learning/Exercises

so that its absolute path is

/content/drive/MyDrive/Colab Notebooks/CPSC 381-581: Machine Learning/Exercises/09_exercise_support_vector_machines.ipynb

In this exercise, we will optimize a perceptron, logistic regression, and support vector machine on 4 datasets. We can compare classification accuracy across all the datasets to see which method is the best.

Submission:

- 1. Implement all TODOs in the code blocks below.
- 2. Report your validation scores for each method averaged over 10 trials.

***** Mean accuracy across 10 trials *****

Perceptron: 0.7937024853801169

Logistic Regression: 0.9523538011695907 Support Vector Machine: 0.9668019005847952

3. List any collaborators.

None

Import packages

```
In []: import numpy as np
import sklearn.datasets as skdata
import sklearn.metrics as skmetrics
```

```
from sklearn.svm import SVC
from sklearn.linear_model import Perceptron
from sklearn.linear_model import LogisticRegression
from matplotlib import pyplot as plt
import warnings

warnings.filterwarnings(action='ignore')
np.random.seed = 1
```

Visualize the decision boundary and support vectors of different classifiers

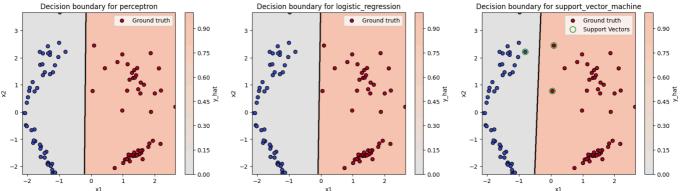
```
In [ ]:
        # Generate synthetic data
        X, y = skdata.make_classification(
            n_features=2,
            n_classes=2,
            n_redundant=0,
            n_clusters_per_class=2,
            n_samples=100,
            class_sep=1.5,
            random_state=1)
        methods = [
            'perceptron',
            'logistic_regression',
             'support_vector_machine'
        # TODO: Create figure with figsize=(20, 5)
        fig = plt.figure(figsize=(20, 5))
        # TODO: Enumerate through methods with index
        for i, method in enumerate(methods):
            # Instantiate model
            if method == 'perceptron':
                # TODO: Instantiate perceptron model with tolerance of 1e-1 and alpha of 0
                model = Perceptron(tol=1e-1, alpha=0)
            elif method == 'logistic_regression':
                # TODO: Instantiate logistic regression model with tolerance of 1e-1
                model = LogisticRegression(tol=1e-1)
            elif method == 'support_vector_machine':
                # TODO: Instantiate SVC (Support Vector Machine Classifier) with tolerance of
                model = SVC(tol=1e-1, C=1e10, kernel='linear')
            else:
                raise ValueError('Unsupported method: {}'.format(method))
            # TODO: Train the model
            model.fit(X, y)
            # TODO: Get x1_min and x1_max (0-th dimension), and x2_min and x2_max (1-st dimen
            x1_{min}, x1_{max} = min(X[:, 0]), max(X[:, 0])
            x2_{min}, x2_{max} = min(X[:, 1]), max(X[:, 1])
            # TODO: Create 2 linspaces: one from x1_min to x1_max and the other from x1_min t
            x1_{linspace} = np.linspace(x1_min, x1_max, 500)
            x2_{linspace} = np.linspace(x2_min, x2_max, 500)
            # TODO: Create meshgrid for x1 and x2 using linspaces
```

```
x1, x2 = np.meshgrid(x1_linspace, x2_linspace)
    # TODO: Predict values for every point in meshgrid
    all_Xs = np.c_[x1.ravel(), x2.ravel()]
    y_hat = model.predict(all_Xs)
    # TODO: Reshape y_hat to x1 or x2's shape
    y hat = y hat.reshape(x1.shape)
    # TODO: Instantiate axis for subplot of a 1 x 3 figure
    ax = fig.add_subplot(1, 3, i + 1)
    # TODO: Plot Contour for predictions with levels=20, cmap='coolwarm', alpha=0.8,
    contour = ax.contourf(x1, x2, y_hat, levels=20, cmap='coolwarm', alpha=0.8, vmin=
    # TODO: Create colorbar for contour on axis and set its label to 'y_hat'
    cbar = fig.colorbar(contour, ax=ax, label='y_hat')
    # TODO: Plot decision boundary using levels=[0], colors='black', linewidths=2
    decision_boundary = ax.contour(x1, x2, y_hat, levels=[0], colors='black', linewid
    # TODO: Create scatter plot for X and set its color to y with edgecolor='black',
    ax.scatter(X[:, 0], X[:, 1], c=y, edgecolor='black', cmap='coolwarm', label='Grou
    # TODO: If support vector machine
    # Create scatter plot of support vectors with s=100, facecolors='none', edgecolor
    if method == 'support_vector_machine':
        ax.scatter(model.support_vectors_[:, 0], model.support_vectors_[:, 1], s=100,
    # TODO: Set title to ''Decision boundary for {}'
    ax.set title('Decision boundary for {}'.format(method))
    # TODO: Set xlabel to 'x1'
    ax.set_xlabel('x1')
    # TODO: Set ylabel to 'x2'
    ax.set_ylabel('x2')
    # TODO: Set legend with loc='upper right'
    ax.legend(loc='upper right')
plt.show()
   Decision boundary for perceptron
                                  Decision boundary for logistic_regression
                                                                Decision boundary for support_vector_machine

    Ground truth

    Ground truth

                                                                               Ground truth
                                                         0.90
                                                                                         0.90
                                                         0.75
                                                                                         0.75
                         0.60
                                                         0.60
                                                                                         0.60
```



Load datasets

```
In []: # Load datasets
    datasets = [
        skdata.load_iris(),
        skdata.load_breast_cancer(),
        skdata.load_digits(),
        skdata.load_wine()
```

```
dataset_names = [
    'Iris',
    'Breast cancer',
    'Digits',
    'Wine'
]
```

Compare Perceptron, Logistic Regression, and Support Vector Machines across all datasets

```
# Define lists to hold validation scores across trials
In [ ]:
        scores_val = {
            'perceptron' : [],
            'logistic_regression': [],
            'support_vector_machine' : []
        }
        # Perform 10 trials of experiments
        n_{trial} = 10
        for n in range(n_trial):
            print('***** TRIAL {} *****\n'.format(n))
            # Zip up all dataset options
            dataset options = zip(
                datasets,
                dataset names)
            for dataset, dataset_name in dataset_options:
                Create the training and validation splits
                X = dataset.data
                y = dataset.target
                print('Preprocessing the {} dataset ({} samples, {} feature dimensions)'.form
                # Shuffle the dataset based on sample indices
                shuffled_indices = np.random.permutation(X.shape[0])
                # Choose the first 80% as training set and the next 20% as validation
                train_split_idx = int(0.80 * X.shape[0])
                train_indices = shuffled_indices[0:train_split_idx]
                val_indices = shuffled_indices[train_split_idx:]
                # Select the examples from X and y to construct our training and validation s
                X_train, y_train = X[train_indices, :], y[train_indices]
                X_val, y_val = X[val_indices, :], y[val_indices]
                for method in ['perceptron', 'logistic_regression', 'support_vector_machine']
                    print('***** Experiments on the {} dataset using {} model *****'.format(
                        dataset_name,
                        method))
                    # Instantiate model
                    if method == 'perceptron':
                        # TODO: Instantiate perceptron model with tolerance of 1e-1 and alpha
                        model = Perceptron(tol=1e-1, alpha=0)
```

```
elif method == 'logistic_regression':
                # TODO: Instantiate logistic regression model with tolerance of 1e-1
                model = LogisticRegression(tol=1e-1)
            elif method == 'support_vector_machine':
                # TODO: Instantiate SVC (Support Vector Machine Classifier) with tole
                model = SVC(tol=1e-1, C=1, kernel='linear')
            else:
                raise ValueError('Unsupported method: {}'.format(method))
            # TODO: Train the model
            model.fit(X_train, y_train)
            # TODO: Score model using mean accuracy on training set
            predictions train = skmetrics.accuracy score(y train, model.predict(X tra
            score_train = skmetrics.accuracy_score(y_train, model.predict(X_train))
            print('Training set mean accuracy: {:.4f}'.format(score_train))
            # TODO: Score model using mean accuracy validation set
            predictions_val = skmetrics.accuracy_score(y_val, model.predict(X_val))
            score_val = skmetrics.accuracy_score(y_val, model.predict(X_val))
            print('Validation set mean accuracy: {:.4f}'.format(score val))
            # TODO: Append score to validation scores for the given method
            scores_val[method].append(score_val)
        print('')
# TODO: Compute mean over trials for each method
mean_scores_val_perceptron = np.mean(scores_val['perceptron'])
mean_scores_val_logistic = np.mean(scores_val['logistic_regression'])
mean_scores_val_svm = np.mean(scores_val['support_vector_machine'])
print('***** Mean accuracy across {} trials *****'.format(n_trial))
print('Perceptron: {}'.format(mean_scores_val_perceptron))
print('Logistic Regression: {}'.format(mean_scores_val_logistic))
print('Support Vector Machine: {}'.format(mean_scores_val_svm))
```

```
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.9500
Validation set mean accuracy: 0.9333
***** Experiments on the Iris dataset using logistic regression model *****
Training set mean accuracy: 0.9667
Validation set mean accuracy: 1.0000
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9833
Validation set mean accuracy: 1.0000
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.8989
Validation set mean accuracy: 0.9561
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9231
Validation set mean accuracy: 0.9737
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9604
Validation set mean accuracy: 0.9825
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9784
Validation set mean accuracy: 0.9472
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9603
Validation set mean accuracy: 0.9528
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9694
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.6408
Validation set mean accuracy: 0.7500
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9225
Validation set mean accuracy: 1.0000
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 0.9930
Validation set mean accuracy: 0.9722
**** TRIAL 1 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.9417
Validation set mean accuracy: 0.9333
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9583
Validation set mean accuracy: 0.9000
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9917
Validation set mean accuracy: 0.9333
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.8945
Validation set mean accuracy: 0.9123
```

***** Experiments on the Breast cancer dataset using logistic_regression model ***** Training set mean accuracy: 0.9275

```
Validation set mean accuracy: 0.9561
***** Experiments on the Breast cancer dataset using support vector machine model ****
Training set mean accuracy: 0.9670
Validation set mean accuracy: 0.9561
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9673
Validation set mean accuracy: 0.9694
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9569
Validation set mean accuracy: 0.9694
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 1.0000
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.6479
Validation set mean accuracy: 0.6667
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9296
Validation set mean accuracy: 0.9444
***** Experiments on the Wine dataset using support vector machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9444
**** TRIAL 2 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.6583
Validation set mean accuracy: 0.7000
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9667
Validation set mean accuracy: 1.0000
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9833
Validation set mean accuracy: 1.0000
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.8835
Validation set mean accuracy: 0.9386
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9275
Validation set mean accuracy: 0.9561
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9582
Validation set mean accuracy: 0.9737
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9631
Validation set mean accuracy: 0.9528
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9645
Validation set mean accuracy: 0.9500
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9722
```

Preprocessing the Wine dataset (178 samples, 13 feature dimensions) ***** Experiments on the Wine dataset using perceptron model *****

```
Training set mean accuracy: 0.5493
Validation set mean accuracy: 0.6111
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9648
Validation set mean accuracy: 0.9722
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 0.9930
Validation set mean accuracy: 1.0000
**** TRIAL 3 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.4750
Validation set mean accuracy: 0.6000
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9667
Validation set mean accuracy: 1.0000
***** Experiments on the Iris dataset using support vector machine model *****
Training set mean accuracy: 0.9750
Validation set mean accuracy: 1.0000
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.9055
Validation set mean accuracy: 0.8947
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9341
Validation set mean accuracy: 0.9123
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9758
Validation set mean accuracy: 0.9211
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9638
Validation set mean accuracy: 0.9250
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9659
Validation set mean accuracy: 0.9444
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9750
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.4014
Validation set mean accuracy: 0.3889
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9789
Validation set mean accuracy: 0.9167
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.8889
**** TRIAL 4 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.6833
Validation set mean accuracy: 0.6333
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9667
Validation set mean accuracy: 0.9667
***** Experiments on the Iris dataset using support_vector_machine model *****
```

Training set mean accuracy: 0.9833
Validation set mean accuracy: 1.0000

Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions) ***** Experiments on the Breast cancer dataset using perceptron model *****

Training set mean accuracy: 0.7824 Validation set mean accuracy: 0.7895

***** Experiments on the Breast cancer dataset using logistic_regression model *****

Training set mean accuracy: 0.9253 Validation set mean accuracy: 0.9474

***** Experiments on the Breast cancer dataset using support_vector_machine model ****

*

Training set mean accuracy: 0.9714 Validation set mean accuracy: 0.9386

Preprocessing the Digits dataset (1797 samples, 64 feature dimensions) ***** Experiments on the Digits dataset using perceptron model *****

Training set mean accuracy: 0.9548 Validation set mean accuracy: 0.9500

***** Experiments on the Digits dataset using logistic_regression model *****

Training set mean accuracy: 0.9659 Validation set mean accuracy: 0.9667

***** Experiments on the Digits dataset using support_vector_machine model *****

Training set mean accuracy: 1.0000 Validation set mean accuracy: 0.9861

Preprocessing the Wine dataset (178 samples, 13 feature dimensions) ***** Experiments on the Wine dataset using perceptron model *****

Training set mean accuracy: 0.6338 Validation set mean accuracy: 0.6111

***** Experiments on the Wine dataset using logistic_regression model *****

Training set mean accuracy: 0.9648 Validation set mean accuracy: 0.9722

***** Experiments on the Wine dataset using support_vector_machine model *****

Training set mean accuracy: 1.0000 Validation set mean accuracy: 0.9722

**** TRIAL 5 ****

Preprocessing the Iris dataset (150 samples, 4 feature dimensions) ***** Experiments on the Iris dataset using perceptron model *****

Training set mean accuracy: 0.9500 Validation set mean accuracy: 1.0000

***** Experiments on the Iris dataset using logistic_regression model *****

Training set mean accuracy: 0.9667 Validation set mean accuracy: 1.0000

***** Experiments on the Iris dataset using support_vector_machine model *****

Training set mean accuracy: 0.9833 Validation set mean accuracy: 0.9667

Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions) ***** Experiments on the Breast cancer dataset using perceptron model *****

Training set mean accuracy: 0.9253 Validation set mean accuracy: 0.8860

***** Experiments on the Breast cancer dataset using logistic_regression model *****

Training set mean accuracy: 0.9297 Validation set mean accuracy: 0.9298

 $***** \ \texttt{Experiments on the Breast cancer dataset using support_vector_machine model} \ ****$

*

Training set mean accuracy: 0.9604 Validation set mean accuracy: 0.9561

Preprocessing the Digits dataset (1797 samples, 64 feature dimensions) ***** Experiments on the Digits dataset using perceptron model *****

Training set mean accuracy: 0.9722 Validation set mean accuracy: 0.9528

```
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9576
Validation set mean accuracy: 0.9611
***** Experiments on the Digits dataset using support vector machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9861
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.6549
Validation set mean accuracy: 0.6111
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9577
Validation set mean accuracy: 0.9167
***** Experiments on the Wine dataset using support vector machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9444
**** TRIAL 6 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.7500
Validation set mean accuracy: 0.8000
***** Experiments on the Iris dataset using logistic regression model *****
Training set mean accuracy: 0.9750
Validation set mean accuracy: 0.9667
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9833
Validation set mean accuracy: 0.9667
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.8374
Validation set mean accuracy: 0.7193
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9319
Validation set mean accuracy: 0.9386
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9736
Validation set mean accuracy: 0.9298
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9687
Validation set mean accuracy: 0.9444
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9652
Validation set mean accuracy: 0.9583
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9722
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.5704
Validation set mean accuracy: 0.6944
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9718
Validation set mean accuracy: 0.8611
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.8889
```

**** TRIAL 7 ****

```
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.6500
Validation set mean accuracy: 0.4000
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9750
Validation set mean accuracy: 0.9667
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9750
Validation set mean accuracy: 1.0000
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.9143
Validation set mean accuracy: 0.9298
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9363
Validation set mean accuracy: 0.9298
**** Experiments on the Breast cancer dataset using support vector machine model ****
Training set mean accuracy: 0.9648
Validation set mean accuracy: 0.9737
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9645
Validation set mean accuracy: 0.9694
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9666
Validation set mean accuracy: 0.9750
***** Experiments on the Digits dataset using support vector machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9833
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.5704
Validation set mean accuracy: 0.7222
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9225
Validation set mean accuracy: 0.9444
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 0.9930
Validation set mean accuracy: 1.0000
**** TRIAL 8 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.4000
Validation set mean accuracy: 0.4667
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9833
Validation set mean accuracy: 0.9000
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9833
Validation set mean accuracy: 1.0000
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.8747
Validation set mean accuracy: 0.9035
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9319
```

Validation set mean accuracy: 0.9386

```
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9692
Validation set mean accuracy: 0.9386
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9798
Validation set mean accuracy: 0.9556
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9673
Validation set mean accuracy: 0.9417
***** Experiments on the Digits dataset using support vector machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.9639
Preprocessing the Wine dataset (178 samples, 13 feature dimensions)
***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.6338
Validation set mean accuracy: 0.6111
***** Experiments on the Wine dataset using logistic_regression model *****
Training set mean accuracy: 0.9507
Validation set mean accuracy: 0.9444
***** Experiments on the Wine dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
Validation set mean accuracy: 0.8889
**** TRIAL 9 ****
Preprocessing the Iris dataset (150 samples, 4 feature dimensions)
***** Experiments on the Iris dataset using perceptron model *****
Training set mean accuracy: 0.6583
Validation set mean accuracy: 0.7000
***** Experiments on the Iris dataset using logistic_regression model *****
Training set mean accuracy: 0.9750
Validation set mean accuracy: 0.9667
***** Experiments on the Iris dataset using support_vector_machine model *****
Training set mean accuracy: 0.9917
Validation set mean accuracy: 0.9667
Preprocessing the Breast cancer dataset (569 samples, 30 feature dimensions)
***** Experiments on the Breast cancer dataset using perceptron model *****
Training set mean accuracy: 0.9187
Validation set mean accuracy: 0.9211
***** Experiments on the Breast cancer dataset using logistic_regression model *****
Training set mean accuracy: 0.9275
Validation set mean accuracy: 0.9561
***** Experiments on the Breast cancer dataset using support_vector_machine model ****
Training set mean accuracy: 0.9604
Validation set mean accuracy: 0.9825
Preprocessing the Digits dataset (1797 samples, 64 feature dimensions)
***** Experiments on the Digits dataset using perceptron model *****
Training set mean accuracy: 0.9562
Validation set mean accuracy: 0.9417
***** Experiments on the Digits dataset using logistic_regression model *****
Training set mean accuracy: 0.9555
Validation set mean accuracy: 0.9528
***** Experiments on the Digits dataset using support_vector_machine model *****
Training set mean accuracy: 1.0000
```

Preprocessing the Wine dataset (178 samples, 13 feature dimensions) ***** Experiments on the Wine dataset using perceptron model *****
Training set mean accuracy: 0.6831

Validation set mean accuracy: 0.9778

Validation set mean accuracy: 0.5556 ***** Experiments on the Wine dataset using logistic_regression model *****

Training set mean accuracy: 0.9648 Validation set mean accuracy: 0.9444

***** Experiments on the Wine dataset using support_vector_machine model *****

Training set mean accuracy: 0.9930 Validation set mean accuracy: 1.0000

**** Mean accuracy across 10 trials ****

Perceptron: 0.7937024853801169

Logistic Regression: 0.9523538011695907 Support Vector Machine: 0.9668019005847952