
Discriminant Analysis

Table of Contents

Load Fisher's Iris Data	1
Apply Linear LDA	2
Visualize the classification	2
Visualize the decision boundary	3
Apply Quadratic DA	4
Visualize the decision boundary	5
Classification Error	6
Confusion Matrix	7
Discriminant Analysis of More than 2 Groups	9
Visualize prediction of hypothetical samples	10
Evaluate prediction of training samples	11

Demonstrates matlab's discriminant analysis functions. by Ahmet Sacan.

Load Fisher's Iris Data

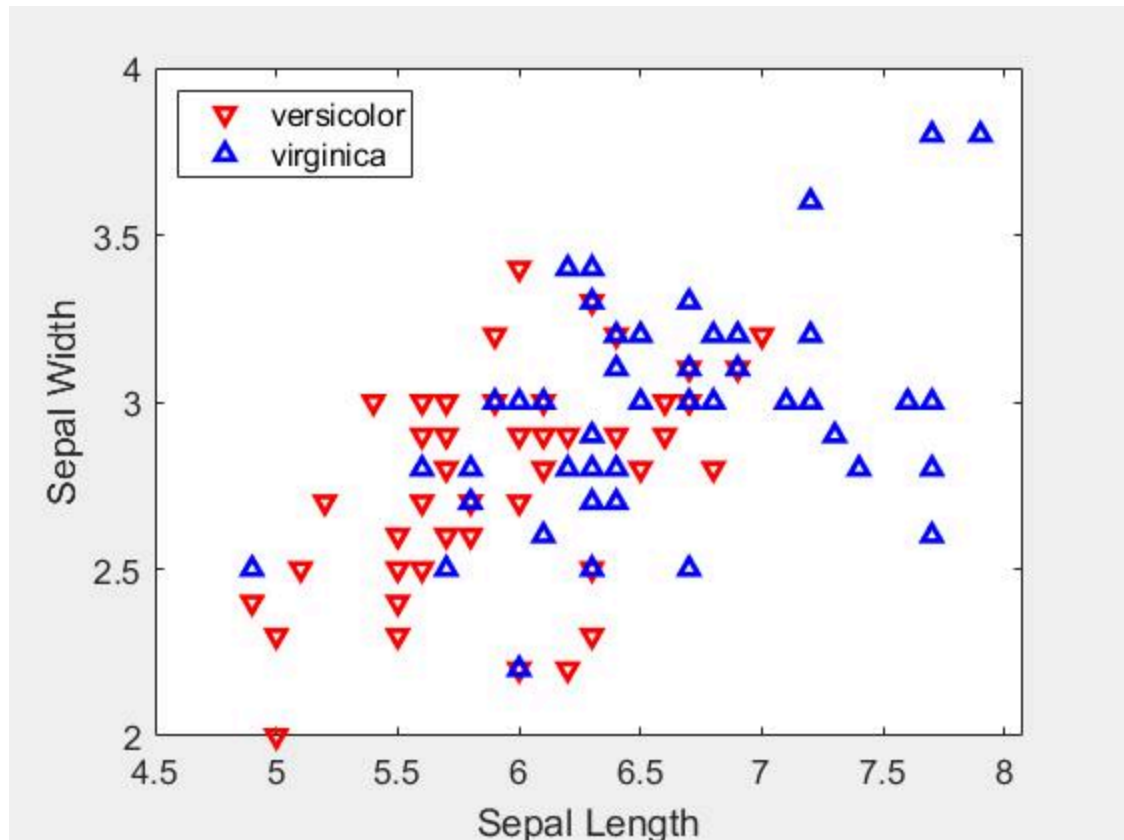
Fisher's iris data consists of measurements on the sepal length, sepal width, petal length, and petal width for 150 iris flower specimens. There are 50 specimens from each of three species: setosa, versicolor, and virginica. See more information at: https://en.wikipedia.org/wiki/Iris_flower_data_set

```
load fisheriris

% We'll skip rows 1..50 which are the 'setosa' species. and use the
% remaining rows, which are versicolor and virginica species.
sepalength = meas(51:end,1);
sepalwidth = meas(51:end,2);
group = species(51:end);

% Let's see how the how the sepal measurements (length and width)
% differ
% between species.
h = gscatter(sepalength,sepalwidth,group,'rb','v^',[],'off');
set(h,'LineWidth',2)
xlabel('Sepal Length'); ylabel('Sepal Width');
legend('versicolor','virginica','Location','NW');

%we'll repeat the visualization of the dataset over and over, so let's
% make
% a function for it.
visualize2classdata=@()
{set(gscatter(sepalength,sepalwidth,group,'rb','v^',
[],'off'),'LineWidth',2) xlabel('Sepal Length') ylabel('Sepal Width')
legend('Fisher versicolor','Fisher virginica','Location','NW')};
```



Apply Linear LDA

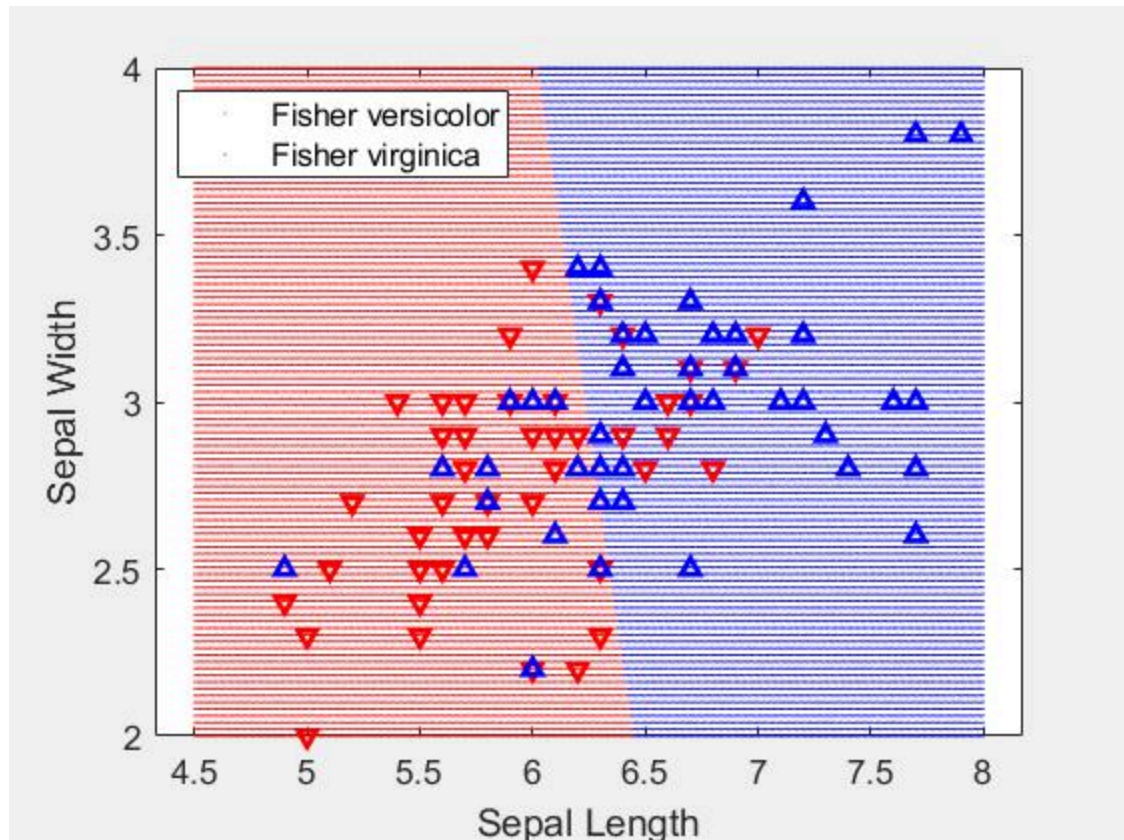
Build/Train the model from the data. We can later apply the model to test cases.

```
mdl = fitcdiscr([sepalength
    sepalwidth],group,'discrimtype','linear');
```

Visualize the classification

We'll classify a meshgrid of sepalength and sepalwidth values into groups and visualize their classification. This is not an ideal way of showing which regions belong to which class, but it works. [L W] are the [sepalength sepalwidth] of our "test" cases.

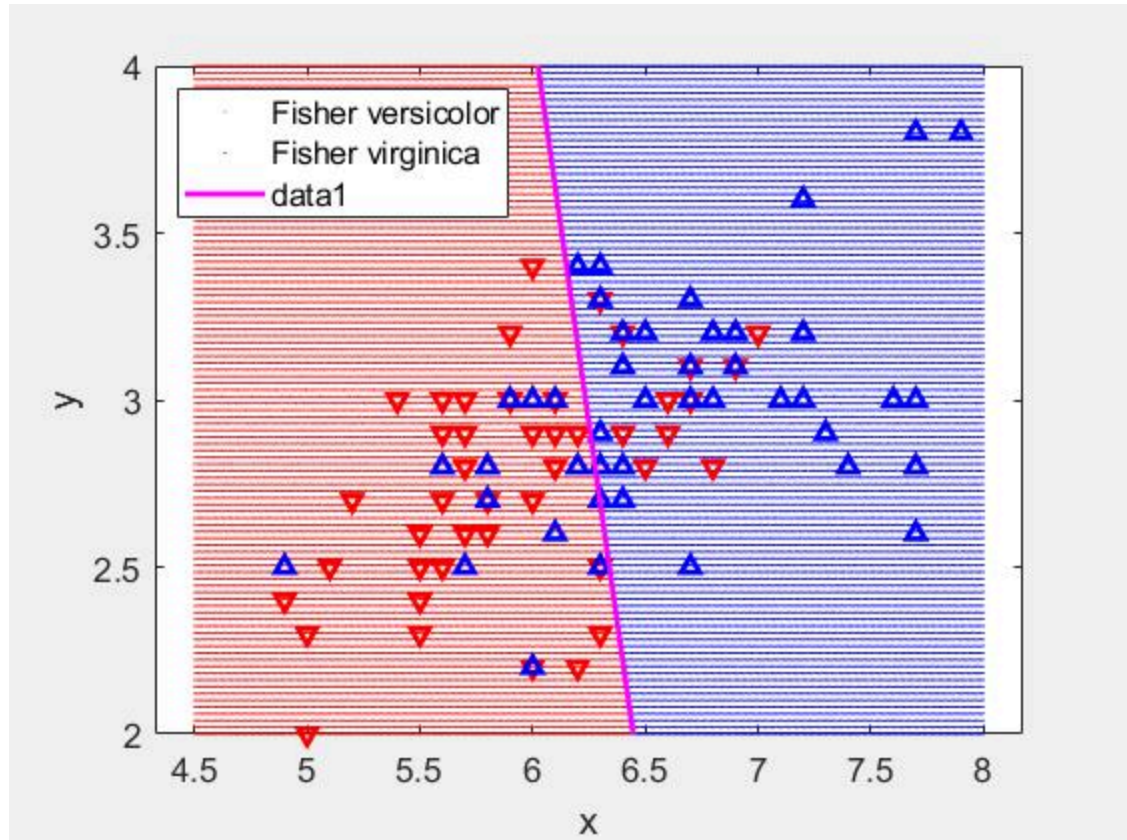
```
[L,W] = meshgrid(linspace(4.5,8,500),linspace(2,4,100)); L=L(:);
W=W(:);
pred = mdl.predict([L W]);
h=gscatter(L,W,pred,'rb','.',1,'off');
set(h,'LineWidth',2,'MarkerSize',2)
hold on; visualize2classdata(); hold off;
```



Visualize the decision boundary

We can actually do better than above. The coefficients we get from the classify function tell us exactly where the classes are separated. We'll use `ezplot()`, which is just an alternative to managing inputs to `plot()` ourselves.

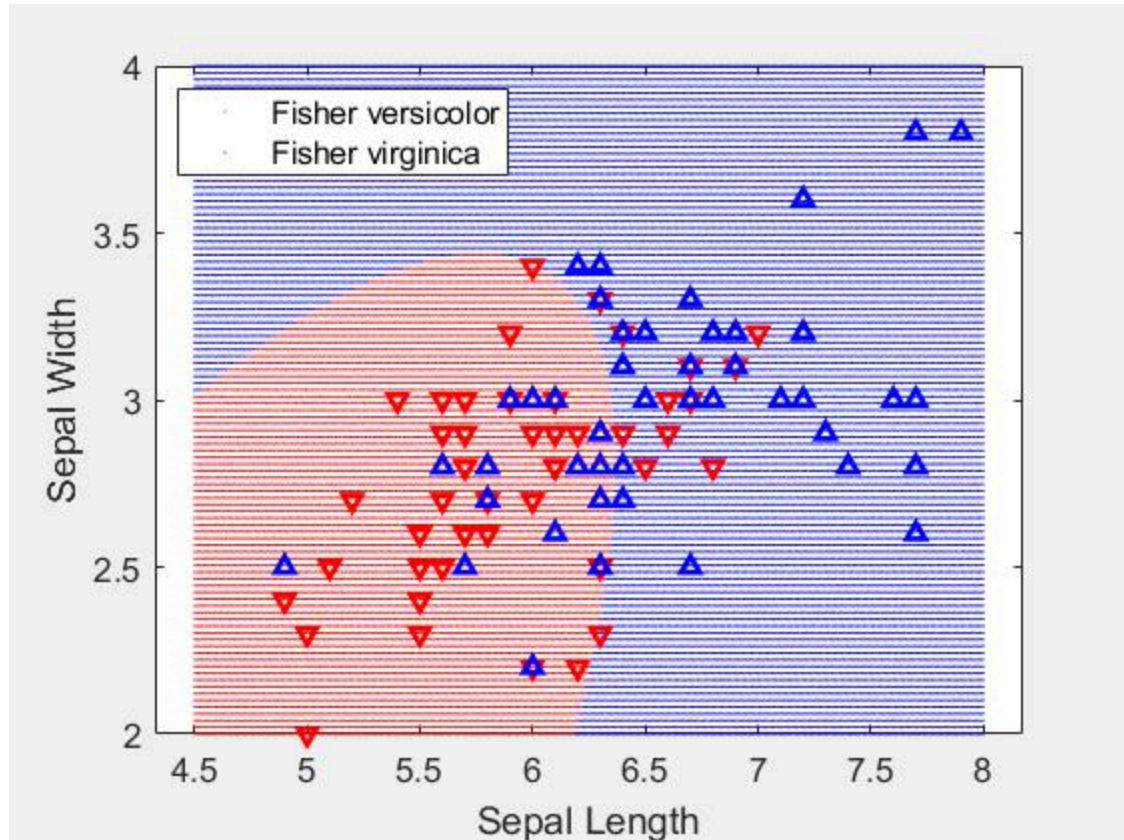
```
hold on;
A = mdl.Coeffs(1,2).Linear;
B = mdl.Coeffs(1,2).Const;
h = ezplot(@(x,y) [x y]*A + B, [4.5 8 2 4]); title('');
set(h, 'Color', 'm', 'LineWidth', 2);
hold off;
```



Apply Quadratic DA

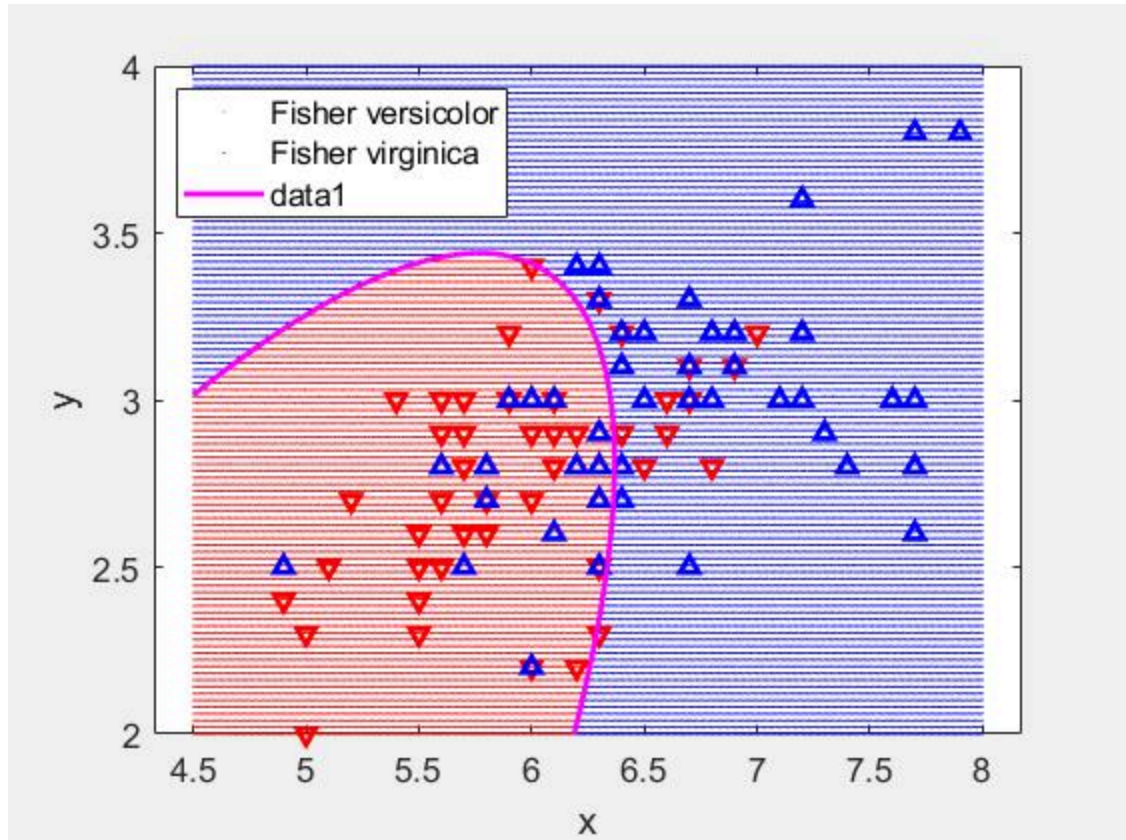
Same as above, but with 'quadratic' option for classify()

```
mdl = fitcdiscr([sepalwidth
    sepalwidth],group,'discrimtype','quadratic');
pred = mdl.predict([L W]);
h=gscatter(L,W,pred,'rb','.',1,'off');
set(h,'LineWidth',2,'MarkerSize',2)
hold on; visualize2classdata(); hold off;
```



Visualize the decision boundary

```
hold on;
A = mdl.Coeffs(1,2).Linear;
B = mdl.Coeffs(1,2).Const;
Q = mdl.Coeffs(1,2).Quadratic;
h = ezplot(@(x,y) sum([x y]*Q).*[x y],2) + [x y]*A + B, [4.5 8 2 4]);
title('');
set(h, 'Color', 'm', 'LineWidth', 2);
hold off
```

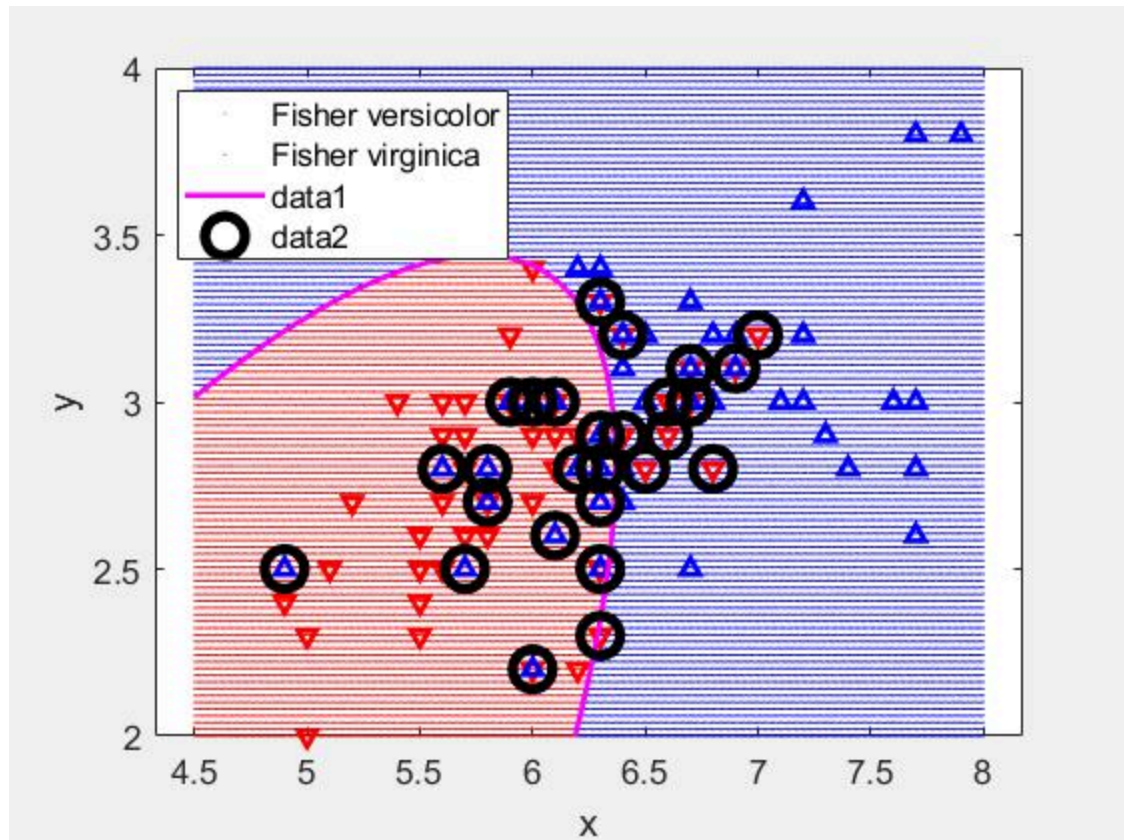
Classification Error

We don't quite know what the correct answers are for [X Y] test cases above. Let's examine how the model performs if we ask it to predict the "training data" whose answers we do know.

```
pred=mdl.predict([sepalength sepalwidth]);
Igood = strcmp(pred,group);
Ibad = ~Igood;
```

show/mark the mis-classified samples on the figure.

```
hold on;
plot(sepalength(Ibad),
     sepalwidth(Ibad), 'ko', 'LineWidth',4, 'MarkerSize',15);
hold off;
```



```
%calculate the misclassification rate (aka error rate)
misclassificationrate = nnz(Ibad) / numel(group)
accuracy = 1-misclassificationrate
```

```
misclassificationrate =
```

```
0.2900
```

```
accuracy =
```

```
0.7100
```

Confusion Matrix

```
[confusionmatrix,matrixlabels] = confusionmat(group,pred)
```

```
confusionmatrix =
```

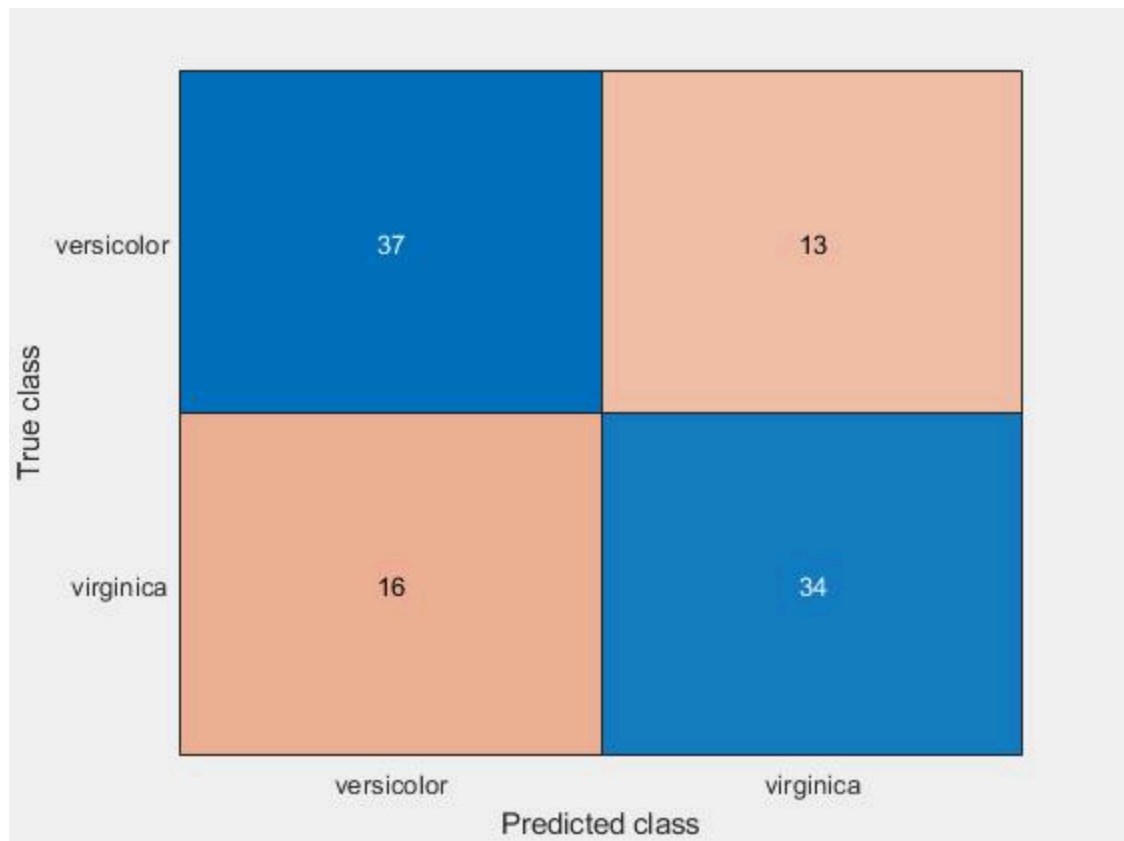
```
37    13
16    34
```

```
matrixlabels =
```

```
2x1 cell array
```

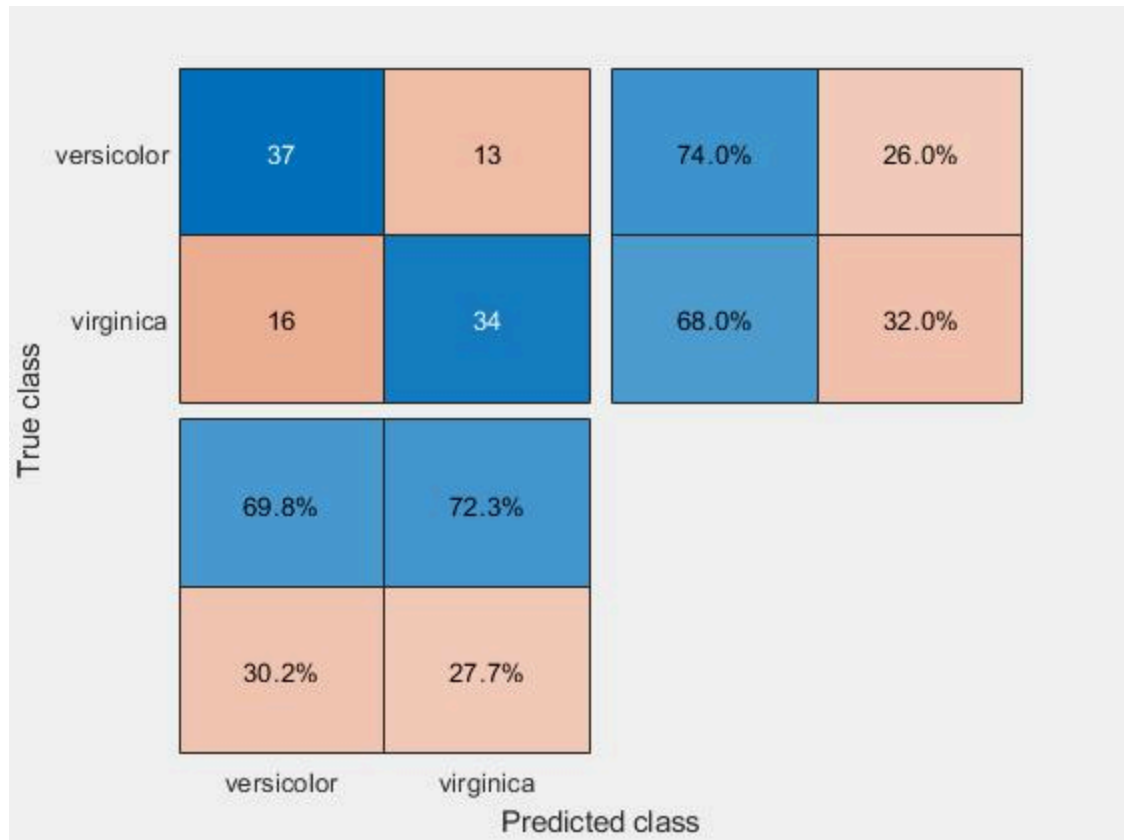
```
{'versicolor'}  
{'virginica' }
```

```
cm = confusionchart(group,pred);
```



```
cm.RowSummary = 'row-normalized';
```

```
cm.ColumnSummary = 'column-normalized';
```

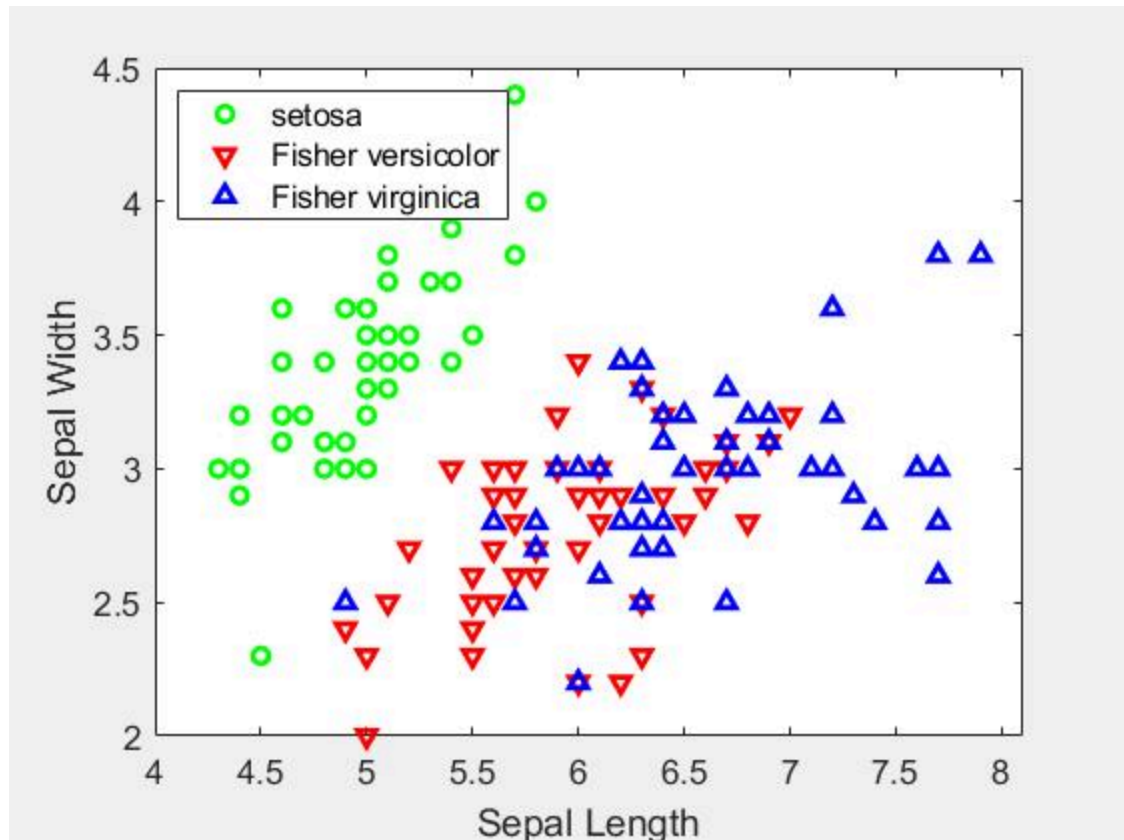
Discriminant Analysis of More than 2 Groups

```

sepalength = meas(:,1);
sepalwidth = meas(:,2);
group = species(:);

h = gscatter(sepalength,sepalwidth,group,'grb','ov^',[],'off');
set(h,'LineWidth',2)
xlabel('Sepal Length'); ylabel('Sepal Width'); legend('setosa','Fisher
versicolor','Fisher virginica','Location','NW');

```



```
mdl = fitcdiscr([sepalLength
    sepalWidth],group,'discrimType','quadratic');
mdl.Coeffs
```

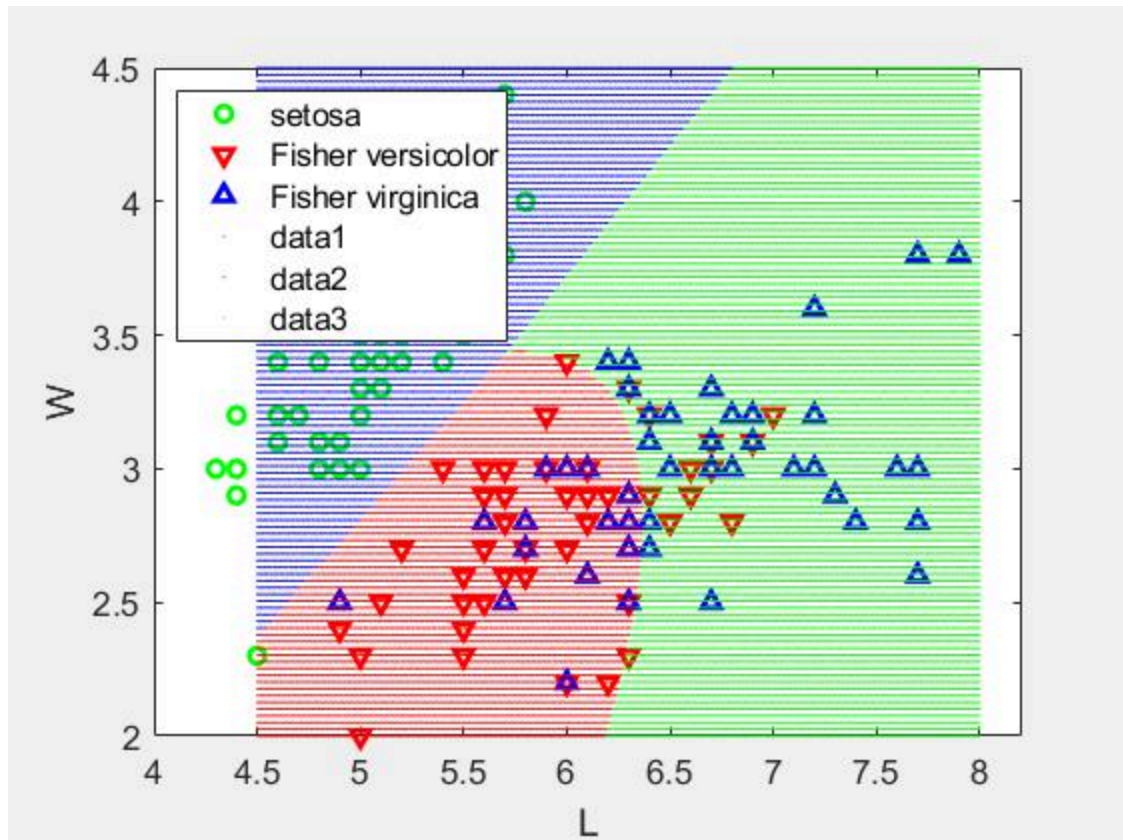
```
ans =
```

```
3x3 struct array with fields:
```

```
DiscrimType
Const
Linear
Quadratic
Class1
Class2
```

Visualize prediction of hypothetical samples

```
[L,W] = meshgrid(linspace(4.5,8,500),linspace(2,4.5,100)); L=L(:);
W=W(:);
pred = mdl.predict([L W]);
hold on;
gscatter(L,W,pred,'rbg','.',1,'off');
hold off
```



Evaluate prediction of training samples

```
pred=mdl.predict([sepalength, sepalwidth]);
Igood = strcmp(pred,group);
accuracy = nnz(Igood) / numel(group)
```

```
accuracy =
```

```
0.8000
```

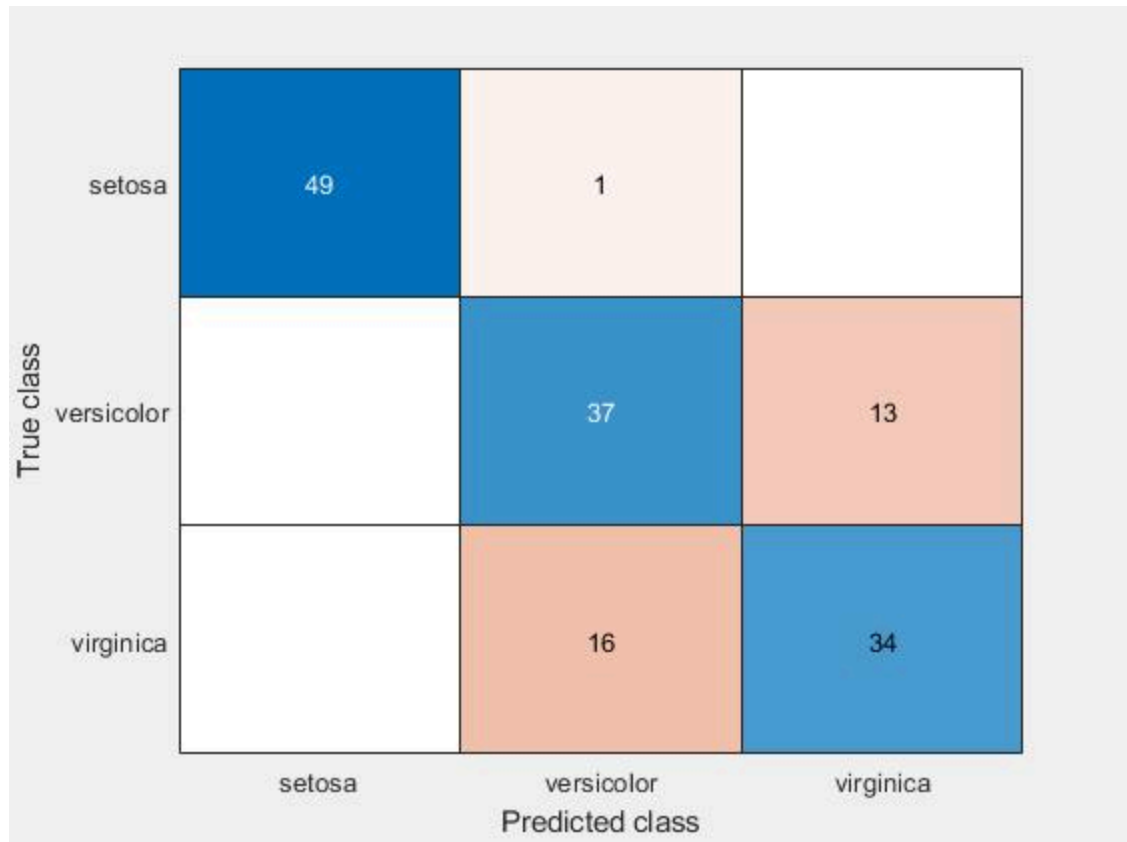
```
confusionchart(group,pred)
```

```
ans =
```

```
ConfusionMatrixChart with properties:
```

```
NormalizedValues: [3×3 double]
ClassLabels: {3×1 cell}
```

```
Use GET to show all properties
```



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