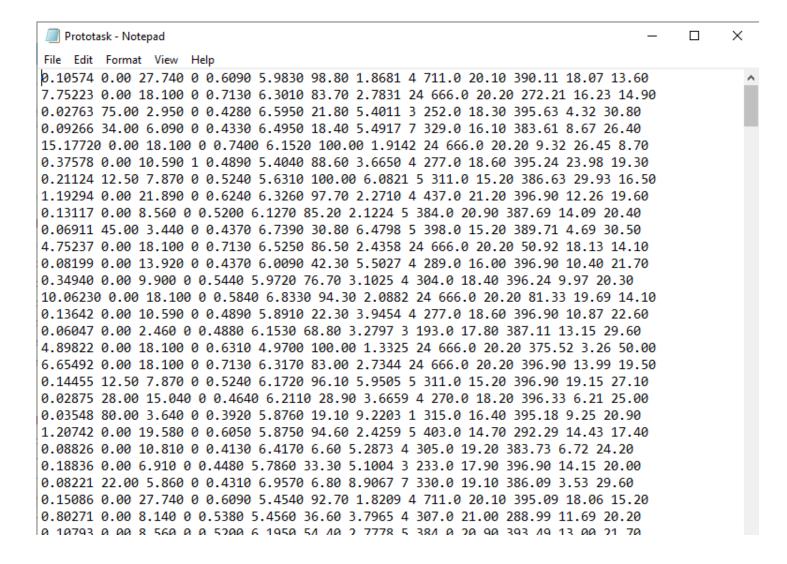
PCA Example

(Boston Data)

Intelligent Data Analysis 2020

Martin Russell

Data (Prototask.data)



- Rows correspond to Boston neighbourhoods
- 506 neighbourhoods in total

Column labels

- 1. CRIM per capita crime rate by town
- 2. ZN proportion of residential land zoned for lots over 25,000 sq.ft.
- 3. INDUS proportion of non-retail business acres per town
- 4. CHAS Charles River dummy variable (1 if tract bounds river; 0 otherwise)
- 5. NOX nitric oxides concentration (parts per 10 million)
- 6. RM average number of rooms per dwelling
- 7. AGE proportion of owner-occupied units built prior to 1940
- 8. DIS weighted distances to five Boston employment centres
- 9. RAD index of accessibility to radial highways
- 10. TAX full-value property-tax rate per \$10,000
- 11. PTRATIO pupil-teacher ratio by town
- 12. B 1000(Bk 0.63)² where Bk is the proportion of blacks by town
- 13. LSTAT % lower status of the population
- 14. MEDV Median value of owner-occupied homes in \$1000's

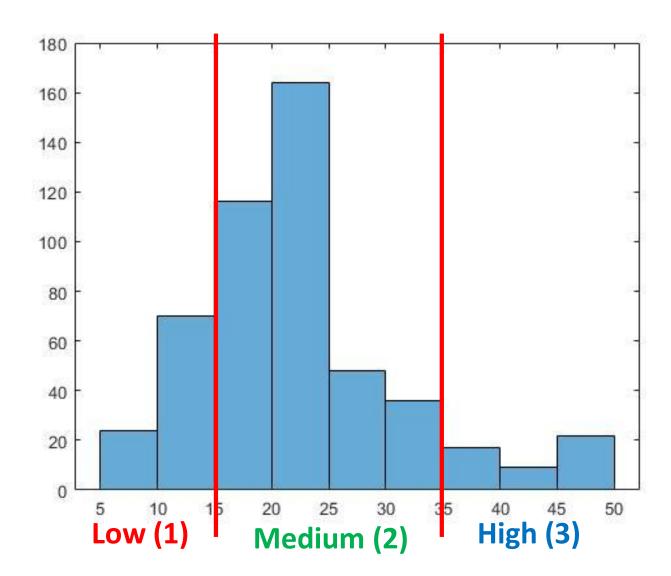
Task

- Two tasks associated with this data set:
 - 1. price can the median value of a home be predicted from the other 13 variables?
 - 2. nox can the nitrous oxide level be predicted from the other 13 variables?
- Use PCA to try to answer these questions

Data pre-processing – Task 1

- Price median value of a home is column 14
- Create two data sets X and L
 - X consists of columns 1 to 13 of the original data
 - This is the data we will apply PCA to
 - L consists of column 14 of the original data
 - This is the 'price' data. We will use this to colour-code (label) the points in the PCA plot
 - First we need to divide the elements of **L** into categories
 - Look at a histogram of the values that L takes

Histogram of the 'price' values



Apply PCA to the data ('price' excluded)

- 1. Calculate the sample mean vector
- 2. Subtract the mean from each data point

```
b = mean(X,1);
```

3. Calculate the covariance matrix of X

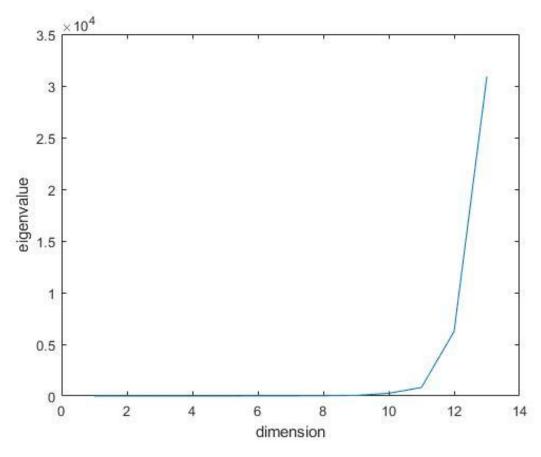
```
C = X'*X/(N-1); (N = number of data points = 506)
```

4. Apply eigenvalue decomposition

$$[U,D] = eig(C);$$

5. Find the two biggest eigenvalues

Eigenvalues



- Biggest eigenvalues are numbers 13 and 12
- Corresponding eigenvalues are 13th and 12th columns of U

Visualise

6. Identify the two principal components

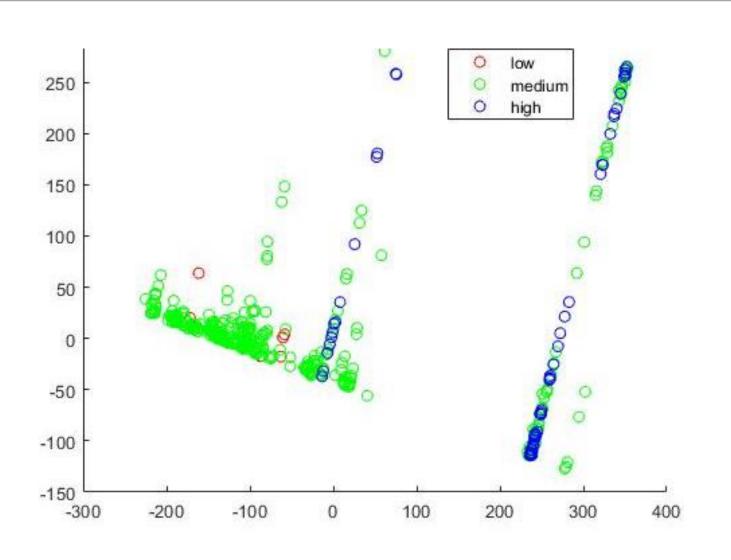
```
e1 = U(:,13);
e2 = U(:,12);
```

7. Project the data points onto the plane

```
x1 = X*e1;
x2 = X*e2;
```

8. Plot the data

```
scatter(x1(L1==1),x2(L1==1),'r');
scatter(x1(L1==2),x2(L1==2),'g');
scatter(x1(L1==3),x2(L1==3),'b');
```



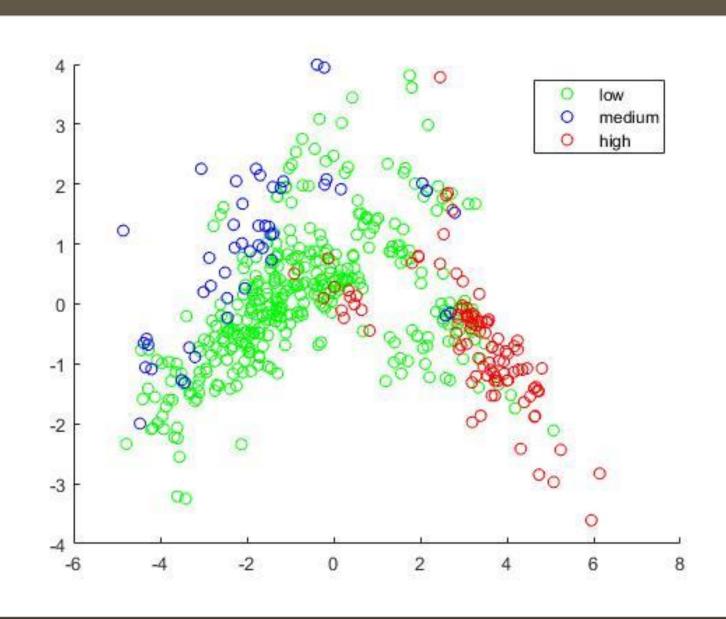
Basic PCA plot ('price' colour coded)

Comments on Basic PCA plot

- Not very informative
- May be due to the difference in the dynamic ranges of the numbers in the different columns
- Normalise each column (for example to have standard deviation 1)

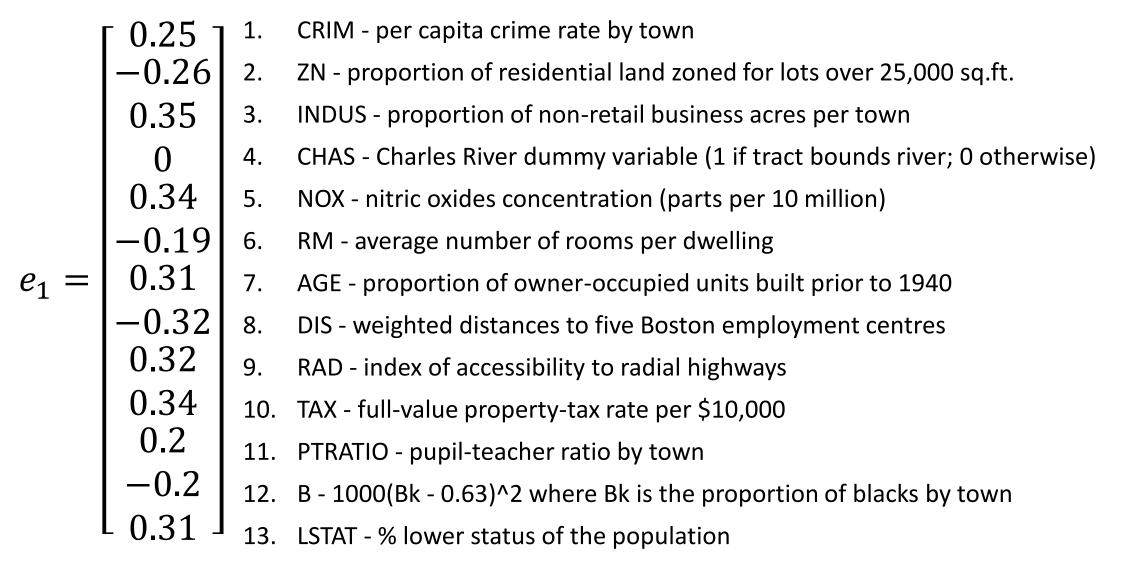
```
m = mean(X,1);
v = var(X,1);
X=(X-m)./sqrt(v);
```

Now repeat the process on slides 7, 8, 9 and 10

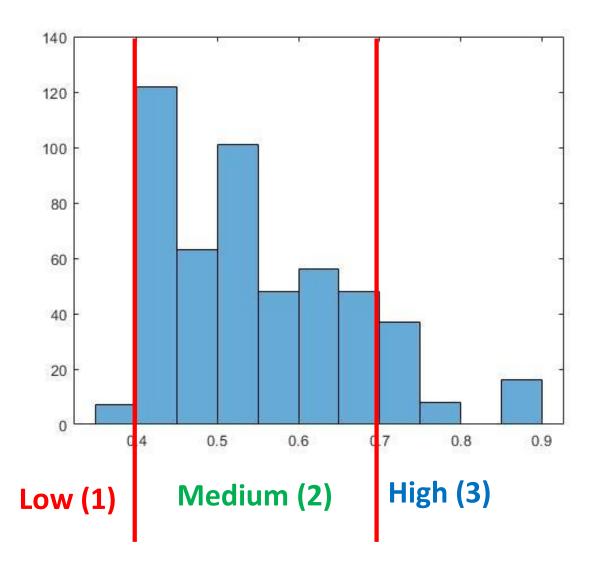


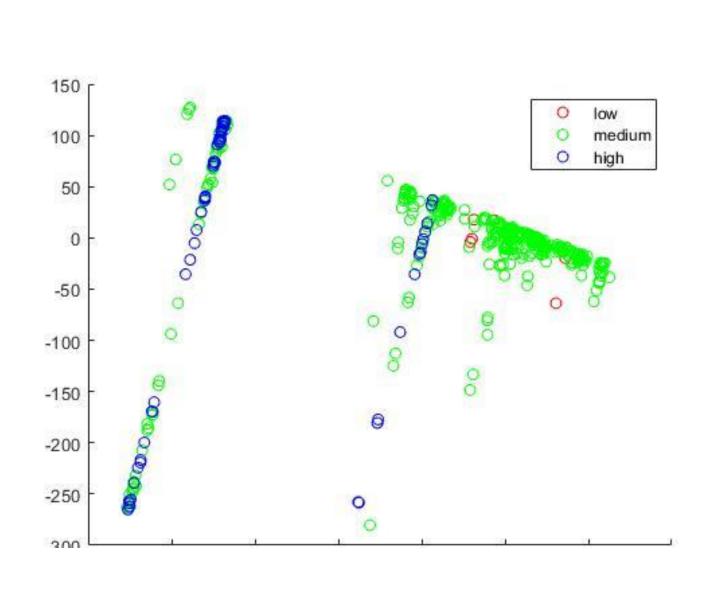
PCA plot – normalised data ('price' colour coded)

1st Principal Component

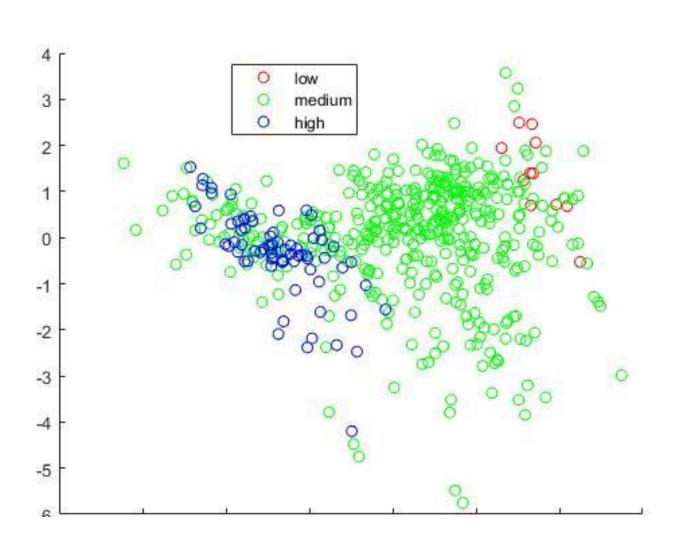


Now repeat for Task 2 – 'nox'





PCA plot – 'nox' – unnormalised



PCA Plot –
'nox' –
variance
normalised

Homework

- Work through the PCA analysis of the boston data with respect to the 'nox' data
- Try to obtain the same figures