# Learning from Data: Visualisation

#### Amos Storkey, School of Informatics

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http://www.anc.ed.ac.uk/~amos/lfd/

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- Presented with new data you should
  - Try to acquire knowledge of how it was created.
  - Visualize the data to see what is in it.
- Visualization is important for understanding what issues there might be with the data, and what forms of assumptions are going to be invalid.
- Visualization is an informal assessment of the data using high level modelling concepts.

• Image: A image:

## Histogram

- What sort of distribution does each attribute of the data have?
- Is it normal?
- Does it have heavy tails?
- Is it skewed?
- Does it cluster?

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### Histogram



- Pretty much Gaussian.
- Tails a bit heavy?

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# Histogram



- Very skewed.
- Some sort of cutoff.
- This is the magnitude of objects in a particular region of sky - more faint objects than very bright ones, but there is a detection limit.



- Plot one attribute against another
- Look at the joint distribution
- Do they appear dependent or independent?
- Are they uniform, peaked, clustered?

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#### **Plots**



- Some dependence
- Positive correlation
- Centred around mean, cannot rule out Gaussianity.

#### **Plots**



- Plot of magnitude (x axis) against area on sky (y axis).
- Definite dependence
- Two clusters, Some outliers

#### **Class Conditional Plots**



- Major elliptical axis (x) versus minor elliptical axis (y) for stars (red) and galaxies (blue).
- Galaxies are more likely to be highly elliptical.
- Two axes definitely related.

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#### **Class Conditional Plots**



- Magnitude (x) versus peak pixel brightness (y)- stars (red), galaxies (blue).
- Stars are more "point like" than galaxies. Higher peak brightness for given magnitude.
- Already have two relevant measures for classifying stars and galaxies.

# **Anomaly Detection**

- Pay attention to outliers, or unusually high peaks in histograms.
- Restrict the data set to those outliers or peaks.
- Attempt to see if there is a potential explanation for them
- You might need to remove outliers to do other analysis

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# **Anomaly Detection**



- Histogram of galaxy orientation
- A few peculiar peaks.
- Restrict data to those within the peak regions.

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# **Anomaly Detection**



- Plot of x-position versus y-position on photograph.
- A long line of points. Wouldn't expect that of galaxies.
- In fact a satellite track which the detection program has mis-recognised.

# **Covariance Visualisation**

- Try to get some idea regarding what variables are dependent.
- Especially appropriate with approximately Gaussian data.
- Do we have positive or negative correlations?
- Is there some structure to the correlations?

#### **Covariance Visualisation**



- Correlation visualisation via Hinton Diagram
- Consecutive pairs correlated.
- Some blockiness.

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- What are the main linear components in the data set. Do they capture any intrinsic concepts?
- Are there any clusters if the data are plotted onto the principal components?

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#### Multi-Attribute Plots

- Turn continuous atrributes into classes. Plot using different colours.
- ► Eg C1 (x > 0, y > 0), C2 (x < 0, y < 0), C3 (x > 0, y < 0), C4 (x < 0, y > 0).
- Plot other attributes conditioned on class.
- Look for class conditional differences.

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#### **Multi-Attribute Plots**



Some variation with colour change.

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#### Summary

- Do not fail to look at your data.
- Real data is always messy. There are going to be things which mess it up.
- Try to identify dependencies.
- How can you reduce the data size: PCA? Ignore components? Ignore data points?

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