Thinking about Data Quit

- 1. ML covered in IAML
 - classification
 - regression
 - dustering
- 2. ML problem
 - large 03 of prev opinion polls and election results, learn a function that can predict a forthcoming election result given current opinion polls
- 3. Numeric appribute 1.1, 2.3, 4.6 Ordinal - Low, Medium, High Categorical - Animal Vegetable, Fish
- 4. Categorical: a=b
- 5. Ordinal: a < b, a=b lequality and ordering)
- 6. Numerical: atb, acb, a=b
- 7. Only whence atts can be normalised
- 8. Purpose of normalization
 -make values of attributes be roughly comparable different
- 9. Normalitation methods
 - converting to range [0.1]
 - converting to zero mean and unit variance
- 10. Outliers can be found by visualising and observing des w/ unusual values stewed distributions can be dealt by taking the log of values A non-monotonic relationship between numerical lordinal values and the class can be dealt by converting values to a variety of overlapping numerical ranges.
- 11. Normalising (e.g. if we scale, centre and destant the image) allows us to use post-normalisation pixel values as features for classification.

 -> depends on the types of images and the classification task.
- 12. For object recognition in images, we want attributes that are: rotation,
 -invariant to irretevant differences in the training images (scale, lighting, etc.)
 -have similar values for images in the same class and different classes
 - spread of values among classes easy to distinguish

- 13. text representation - use the possible words (vocabulary) as the outs and the existence of that word in the document as the value. - word count
- 14. sound organis (speech, music etc.) representation frequencies present in signal
- 15. Accuracy is a good success metric for comparing classifiers if ... the classes are balanced.
- 16. Generative moders dasses
 - ahmost always probabilistic can use unlabeled data

 - Discriminative models decision boundary
 - may/may not be probabilistic can't use unlabeled data