

## Today in MA40189:

- consider an experiment with possible **outcomes**  $\theta \in \Theta$ 
  - chose some **decision**  $d \in \mathcal{D}$  where the consequences depend upon  $\theta$
  - **loss function**  $L(\theta, d)$  measures consequence of decision  $d$  if  $\theta$  occurs
  - want to **minimise loss**: if  $L(\theta, d_1) < L(\theta, d_2)$  we prefer  $d_1$  to  $d_2$
- focus upon **statistical decision theory**: consider  $d$  as being a method of inference for  $\theta$
- statistical decision problem:  $[\Theta, \mathcal{D}, \pi(\theta), L(\theta, d)]$ 
  - solve  $[\Theta, \mathcal{D}, f(\theta), L(\theta, d)]$  for **immediate decision**
  - solve  $[\Theta, \mathcal{D}, f(\theta | x), L(\theta, d)]$  for **decision having observed the sample  $x$**
- **Bayes risk**  $\rho^*(\pi)$  minimises expected loss
$$\rho(\pi, d) = \int_{\theta} L(\theta, d)\pi(\theta) d\theta$$
- **Bayes rule**  $d^*$  decision which achieves Bayes risk
- example of **quadratic loss**: Bayes rule is the mean, Bayes risk is the variance