

# Bayes' Formula

## Query problem (lecture)

$$P(R_d|Q) = \frac{P(Q|R_d)P(R_d)}{P(R_d)}$$

- $R_d = 1$  if document  $d$  relevant,  $Q = (q_1, \dots, q_n)$  query.
- Recall: Unigram assumption, so  $P(Q|R_d) = \prod_{i=1}^n P(q_i|R_d)$ .

## Bayes' formula: Schematic

$$P(\text{State}|\text{Observation}) = \frac{P(\text{Observation}|\text{State})P(\text{State})}{P(\text{Observation})}$$

State = Random value to estimate from dependence on observation.

## Components

$$\text{Posterior} = \frac{\text{Likelihood} \times \text{Prior}}{\text{Evidence}}$$

# Meaning of Components

## Components

$$\text{Posterior} = \frac{\text{Likelihood} \times \text{Prior}}{\text{Evidence}}$$

- **Posterior:** The model to compute.
- **Evidence:** Given  $Q$ , evidence is constant, not important for comparison of different  $d$  (same  $Q$ ).
  - ▶ Neglect evidence: For comparison between different  $d$  for same  $Q$  (e.g. retrieve  $d_1$  or  $d_2$ ?). Suitable for ranking.
  - ▶ Required: For absolute scoring (e.g. retrieve all documents with  $P(R_d|Q) > \text{threshold}$ )
- **Likelihood:**  $P(Q|R_d)$  means: How likely would it be to generate the query  $Q$  we have observed, if we knew that  $d$  is relevant?
- **Prior:** Probability of the document being relevant “in general”. May be based on any information about the documents available when model is computed, but *not on the query*.

# Likelihood Idea

## Problem setting

- Given: Two “states”  $y_1, y_2$ , observation  $x$ , likelihood  $P(x|Y)$ .
- Examples of states: Relevance above, model/parameter etc
- Given that  $x$  has been observed, which one should we choose?

## Maximum likelihood principle

Choose the one which makes  $x$  more probable to happen.

## Estimation strategies

- 1 Pick the state that maximizes the likelihood.
- 2 Pick the state that maximizes the posterior.

## Likelihood only vs. full Bayesian formula

Likelihood models generation of observation given state; does not take into account information about state.