

Bayesian decision with discrete probabilities

What we know:

A student visits open days at three different universities. The student knows that that the uni have done the following decisions about making applications to their degrees for students attending their open days.

- University 1 $30\% \rightarrow 0,3$
- University 2 $50\% \rightarrow 0,5$
- University 3 $20\% \rightarrow 0,2$

Further knows the student the likelihood based on the analysis of the student experiences (X) in the open day if its exciting or boring

- Uni 1 Exciting $0,7$
- Uni 2 " $0,6$
- Uni 3 " $0,3$

What we need:

The student would like to know where he shall apply for a degree.

- ① First he wants to know what the probability of a exciting open day is, for all universities.
- ② Help to decide based on posterior probabilities, also show him the risk
- ③ The student asks himself how he could minimize the error (advise)
- ④ The student likes to know what the loss function is, can you show it to him?
- ⑤ Use the loss function to minimize the error rate and advise where he should apply.

1) Formula

$$P(w_i | x) = P(x | w_i) \cdot P(w_i)$$

likelihood #3

Prior probability #1

#4 Posterior Probability

P(x)

Marginal likelihood #2

or evidence

#order
to calculate

2) Fill in what you know example University 1

$$P(w_i | x) = \frac{0,7 \times 0,3}{P(x)}$$

$P(x)$ = Marginal likelihood is missing

3) Formula (we know from week 2 - 18.10) Answer ①

Where in case of two categories the "evidence"

$$P(x) = \sum_{j=1}^{j=3} p(x | w_j) P(w_j) = P(\text{exciting}) = \sum_{\text{univ. 1}}^{\text{univ. 3}} p(\text{exciting}) \cdot p(\text{acceptance})$$

$$p(\text{exciting}) = \begin{bmatrix} 0,7 \times 0,3 \\ 0,6 \times 0,5 \\ 0,3 \times 0,2 \end{bmatrix} = 0,57 \quad p(\text{boring}) \text{ only when two classes} \\ 1 - 0,57 = 0,43$$

4) Fill into the formula (Uni 1)

Answer ②

$$p(w_1 | \text{exciting}) = \frac{0,7 \times 0,3}{0,57}$$

↑
University

$$p(w_2 | \text{exciting}) = \frac{0,6 \times 0,5}{0,57} = 0,37$$

$$p(w_3 | \text{exciting}) = \frac{0,3 \times 0,2}{0,57} = 0,11$$

because there are
two classes
exciting & boring = 1

5) Help to decide

- we know that Decision given the posterior probabilities

if $P(w_1|x) > P(w_2|x)$ \Rightarrow True state of nature = w_1

- Therefore we can order:

$$P(u_2|x) > P(u_1|x) > P(u_3|x)$$

- The posterior probability of University is the highest, thus we apply there.

6) Wrong choice ?!

• Whenever we observe a particular x , the probability of error is:

$$P(\text{error}|x) = P(w_1|x) \text{ if we decide } w_2$$

$$P(\text{error}|x) = P(w_2|x) \text{ if we decide } w_1$$

• Therefore

$$P(\text{error}|x) = \min [P(w_1|x), P(w_2|x)]$$

• Transform

$$P(\text{error}|x) = \min [P(u_1|x), P(u_2|x), P(u_3|x)]$$

$$\underline{0, 10} = \min [(0, 27), (0, 53), (0, 10)]$$

The probability of making the wrong choice is at least 10%

(3) In case we want to minimize this error further, we can increase the numbers of samples, increase the dimension of feature space by using more features, thereby changing the likelihood $P(x|m)$ or combination of both

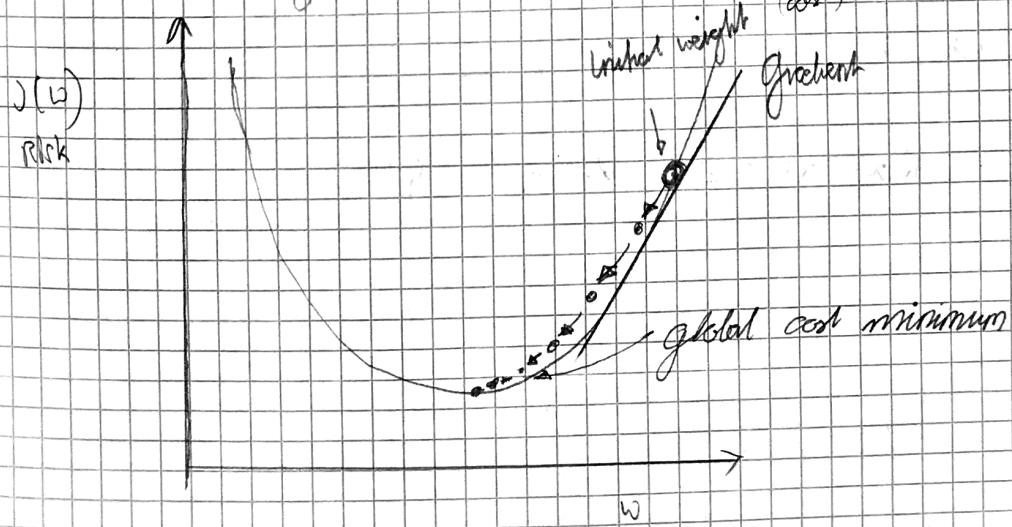
- Use more than one feature
- Use more than two states of nature
- allowing actions that only decide on the state of nature
- Introduce a loss function which is more general than the probability of error.

(4) Introducing the loss function (cost function)

- allowing actions other than classification primarily allows the possibility of rejection
- Refusing to make a decision in close or bad cases!
- The loss function states how costly each action taken is.

The loss or cost function is usually used to find the Global cost minimum. This is an optimization problem and in this context are called Bayes estimator. We can feed the model with more important features such as (costs) in this example and calculate the (local) minimum (global)

General idea of cost function



⑤ Theory

- Let $\{w_1, w_2, w_3\}$ be the set of c states of nature "categories" (Umstände)
- Let $\{\alpha_1, \alpha_2, \alpha_3\}$ be the set of possible actions
- Let $\lambda(\alpha_i | w_j)$ be the loss incurred for taking action α_i when the state of nature is w_j .
- $R = \text{sum of all } R(\alpha_i | x)$ for i

1) You need an glorifier:

$$U_1 = \bar{s} - 4K < \bar{s}(U_2) > \bar{s}(U_3)$$

$$U_1 = -4 \text{ n than } U_2$$

$$U_2 = +1 \text{ n than } U_3$$

2) in 20 years that would mean

	U_1	U_2	U_3
Cost	80	0	20

3) State of actions (apply = Reject) α

4) final Table (from vector)

	U_1	U_2	U_3
apply	80	0	100
reject	60	90	0

5) Formula $R(\alpha_i | x) = \sum_{j=1}^{i+3} \lambda(\alpha_i | w_j) P(w_j | x)$

$$R(\alpha_1 | x) = \begin{pmatrix} 80 \times 0,37 \\ 0 \times 0,57 \\ 100 \times 0,1 \end{pmatrix} = 39,6$$

$$R(\alpha_2 | x) = \begin{pmatrix} 60 \times 0,37 \\ 90 \times 0,57 \\ 0 \times 0,37 \end{pmatrix} = 69,9$$

6) Credit Risk

Select the action a for which $R(\alpha | x)$ is minimum in this case, R is called the 'Bayes' Risk which is the best performance that can be achieved.

Answer: Apply to a place in the university because over 20 years it will cost you more than to reject a university offer.

7) Improve decisions (self made)

	u1	u2	u3
Apply to u1	0	20	40
Apply to u2	80	0	20
Apply to u3	30	30	0

$$R(\alpha_1 | x) = 0 \times 0,37 + 20 \times 0,53 + 40 \times 0,1 = 19,6$$

$$R(\alpha_2 | x) = 80 \times 0,37 + 0 \times 0,53 + 20 \times 0,1 = 31,6$$

$$R(\alpha_3 | x) = 30 \times 0,37 + 30 \times 0,53 + 0 \times 0,1 = 53,5$$

8) According to this, which may sound counter intuitive you should apply to u1 and not to u2.