

Exercise 10

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$$1a) \quad 1 = g(z) + g(-z), \quad g(z) = \frac{1}{1 + e^{-z}}$$

$$\begin{aligned} g(z) + g(-z) &= \frac{1}{1 + e^{-z}} + \frac{1}{1 + e^z} \\ &= \frac{1 + e^z + 1 + e^{-z}}{(1 + e^{-z})(1 + e^z)} \\ &= \frac{2 + e^z + e^{-z}}{1 + e^{-z} + e^z + e^z e^{-z}} \\ &= \frac{2 + e^z + e^{-z}}{2 + e^z + e^{-z}} = 1 // \end{aligned}$$

$$1b) \quad \vec{z} = \vec{w}^T \vec{x} = \begin{pmatrix} \vec{w}^T \\ -\vec{w}^T \end{pmatrix} \cdot \vec{x} \Rightarrow \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} \vec{w}^T \vec{x} \\ -\vec{w}^T \vec{x} \end{pmatrix} \equiv \begin{pmatrix} z \\ -z \end{pmatrix}$$

We want to show that we get the same probabilities for $P(\text{class } i | \vec{w}, \vec{x}) = g(z_i) = [\sigma(\vec{z})]_i$ by rescaling \vec{w} in the softmax function σ .

$$\begin{aligned} \sigma(z_1) &= \frac{e^{z_1}}{e^{z_1} + e^{z_2}} = \frac{e^z}{e^z + e^{-z}} = \frac{1}{1 + e^{-2z}} \\ &= \frac{1}{1 + e^{-2\vec{w}^T \vec{x}}} \stackrel{!}{=} \frac{1}{1 + e^{-\vec{w}'^T \vec{x}}} = g(z) \end{aligned}$$

$$\Rightarrow \vec{w}' = \frac{\vec{w}}{2}$$

So with $\bar{w}' = \bar{w} \cdot \frac{1}{2} = \frac{1}{2} \cdot \begin{pmatrix} w^T \\ -w^T \end{pmatrix}$ we get the same probability-results with the sigmoid and softmax function //

2) a) Joint probability $P(x, y)$ means that we want to know the probability that both events x and y occur together,

Marginal probability $P(x)$ means that we want to know the probability that the event x occurs regardless of the other outcome of y . So

$$P(x) = \sum_y P(x, y)$$

Conditional probability $P(x|y)$ is the probability of the event x occurring under the condition that the event y already happened.

b) We know: $P(a, b) = P(b, a)$

↓ prob. chain rule

$$P(a|b) \cdot P(b) = P(b|a) \cdot P(a)$$

$$\Rightarrow P(a|b) = \frac{P(b|a) \cdot P(a)}{P(b)} = \text{Bayes' eq.} //$$

$$P(b|a) = \frac{P(a|b) \cdot P(b)}{P(a)}$$

c) $P(a|b)$ is the posterior probability, that's what we want to know.

$P(a)$ is the prior-probability. The probability of the event a occurring that we know beforehand. And $P(b|a)$ is the likelihood, so the information of our current situation that we combine with the prior-prob. to get the outcome (posterior-prob.).

d)

$P(R, S)$	$R=0$	$R=1$
$S=0$	0.4	0.2
$S=1$	0.3	0.1

$$\sum_{R \in \{0,1\}} \sum_{S \in \{0,1\}} P(R, S) = P(0,0) + P(0,1) + P(1,0) + P(1,1) \\ = 0.4 + 0.2 + 0.3 + 0.1 \\ = 1 //$$

$$P(R=0) = \sum_S P(0, S) = P(0,0) + P(0,1) = 0.4 + 0.3 = 0.7$$

$$P(R=1) = \sum_S P(1, S) = P(1,0) + P(1,1) = 0.2 + 0.1 = 0.3$$

$$P(S=0) = \sum_R P(R, 0) = P(0,0) + P(1,0) = 0.4 + 0.2 = 0.6$$

$$P(S=1) = \sum_R P(R, 1) = P(0,1) + P(1,1) = 0.3 + 0.1 = 0.4$$

$$P(R|S) = \frac{P(R, S)}{P(S)}$$

$$P(0|0) = \frac{0.4}{0.6} = \frac{2}{3}$$

$$P(0|1) = \frac{0.3}{0.4} = \frac{3}{4}$$

$$P(1|0) = \frac{0.2}{0.6} = \frac{1}{3}$$

$$P(1|1) = \frac{0.1}{0.4} = \frac{1}{4}$$

— sums to 1 ✓

— sums to 1 ✓

$$P(S|R) = \frac{P(S,R)}{P(R)} \quad P(0|0) = \frac{0,4}{0,7} = \frac{4}{7}$$

$$P(0|1) = \frac{0,2}{0,3} = \frac{2}{3}$$

$$P(1|0) = \frac{0,3}{0,7} = \frac{3}{7}$$

$$P(1|1) = \frac{0,1}{0,3} = \frac{1}{3}$$

$R=1, S=1$ with Bayes eq.

$$P(R=1|S=1) = \frac{P(S=1|R=1) \cdot P(R=1)}{P(S=1)} = \frac{\frac{1}{3} \cdot 0,3}{0,4} = \frac{1}{4} \checkmark$$

$$P(S=1|R=1) = \frac{P(R=1|S=1) \cdot P(S=1)}{P(R=1)} = \frac{\frac{1}{4} \cdot 0,4}{0,3} = \frac{1}{3} \checkmark$$

		disease		
		$a=1$	$a=0$	
test	$b=1$	0,0095	0,0495	$P(a=1) = 0,01$ $P(b=1 a=1) = 0,95$
	$b=0$	0,0005	0,9405	$P(b=0 a=0) = 0,95$

Test is positive $b=1$, what is the probability that Joe has the disease? $P(a=1|b=1)$

$$P(a|b) = \frac{P(b|a) \cdot P(a)}{P(b)} = \frac{0,95 \cdot 0,01}{0,059} \approx 16,10\%$$

3)

