

# Brain Inspired Computing - Sheet 7

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## Übung 7

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We define the disease set  $\{P, N\}$  & the test set  $\{p, n\}$  with positive ( $p \& P$ ) and negative ( $n \& N$ ) results.

The sensitivity is then:

$$p(p|P) = 0.95 \rightarrow p(n|P) = 1 - p(p|P) = 0.05$$

The specificity is:

$$p(n|N) = 0.9 \rightarrow p(p|N) = 1 - p(n|N) = 0.1$$

We can use the incidence to express our illness probability:

$$p(P) = \frac{50}{100\,000} \vee \frac{500}{100\,000} \vee \frac{5000}{100\,000}$$

as well as the healthy probability:

$$p(N) = 1 - p(P) = 1 - \frac{50}{100\,000} \vee 1 - \frac{500}{100\,000} \vee 1 - \frac{5000}{100\,000}$$

a) We are interested in the probability  $p(N|n)$

According to the Bayes' theorem,

$$p(N|n) = \frac{p(n|N)p(N)}{p(n)}$$

$$\text{where } p(n) = p(n|N)p(N) + p(n|P)p(P)$$

We obtain:

$$p(N|n) = \frac{p(n|N)p(N)}{p(n|N)p(N) + p(n|P)p(P)}$$

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$$p(N|n) = \frac{p(n|N)p(N)}{p(n|N)p(N) + p(n|P)p(P)}$$

with values:

$$p(N|n) \approx 0.999972 \approx 99.997\% \quad (\text{Incidence } 50)$$

$$\checkmark 0.99972 \approx 99.97\% \quad (\text{Incidence } 500)$$

$$\checkmark 0.9971 \approx 99.71\% \quad (\text{Incidence } 5000)$$

b) we want:  $p(P|p)$ :

$$p(P|p) = \frac{p(p|P)p(P)}{p(p|P)p(P) + p(p|N)p(N)}$$

$$\Rightarrow p(P|p) \approx 0.00473 \approx 0.473\% \quad (\text{Incidence } 50)$$

$$\checkmark 0.04556 \approx 4.556\% \quad (\text{Incidence } 500)$$

$$\checkmark 0.66 \approx 66.67\% \quad (\text{Incidence } 5000)$$

We now set  $p(P) = 0.6 \rightarrow p(N) = 0.4$

and see, that the results of c) & d)

are no longer dependent on the actual incidence:

$$c) p(P|p) = \frac{p(p|P)p(P)}{p(p|P)p(P) + p(p|N)p(N)} \approx 0.9344 \approx 93.44\%$$

$$d) p(P|n) = \frac{p(n|P)p(P)}{p(n|P)p(P) + p(n|N)p(N)} \approx 0.0769 \approx 7.69\%$$

## Exercise 2

(a)

$$p(z_k^{n+1} = 1 \mid z_{\setminus k}) = \frac{p(z_k^{n+1} = 1 \mid z_{\setminus k})}{p(z_k^{n+1} = 0 \mid z_{\setminus k}) + p(z_k^{n+1} = 1 \mid z_{\setminus k})} = \frac{1}{1 + \frac{p(z_k^{n+1}=0 \mid z_{\setminus k})}{p(z_k^{n+1}=1 \mid z_{\setminus k})}} \quad (1)$$

Using the definition of conditional probability

$$p(y|x) = \frac{p(y, x)}{p(x)} \quad (2)$$

one gets that

$$\frac{p(z_k^{n+1} = 0 \mid z_{\setminus k})}{p(z_k^{n+1} = 1 \mid z_{\setminus k})} = \frac{p(z_k^{n+1} = 0, z_{\setminus k})}{p(z_k^{n+1} = 1, z_{\setminus k})} = \frac{p(\mathbf{z})}{p(\mathbf{z}')} \quad (3)$$

where

$$\mathbf{z} = (z_1, \dots, z_{k-1}, 0, z_{k+1}, \dots, z_n) \quad \mathbf{z}' = (z_1, \dots, z_{k-1}, 1, z_{k+1}, \dots, z_n)$$

Using the probability density function

$$p(\mathbf{z}) = \frac{1}{Z} \exp \left[ \frac{1}{2} \mathbf{z}^t W \mathbf{z} + \mathbf{b}^t \mathbf{z} \right]$$

together with the fact that  $\mathbf{z}, \mathbf{z}'$  are equal a part from the term  $k$  and that  $z_k = 0, z'_k = 1$  we have now that

$$\frac{p(\mathbf{z})}{p(\mathbf{z}')} = \exp \left[ \frac{1}{2} \left( \mathbf{z}^t W \mathbf{z} + \mathbf{b}^t \mathbf{z} - \mathbf{z}'^t W \mathbf{z}' - \mathbf{b}^t \mathbf{z}' \right) \right] = \exp \left[ \frac{1}{2} \left( \mathbf{z}^t W \mathbf{z} - \mathbf{z}'^t W \mathbf{z}' - b_k \right) \right]$$

The quadratic form can be worked out as

$$\mathbf{z}^t W \mathbf{z} - \mathbf{z}'^t W \mathbf{z}' = z_i W_{ij} z_j - z'_i W_{ij} z'_j = -z'_k W_{kj} z'_j - z'_i W_{ik} z'_k = -2z'_k W_{kj} z'_j$$

where  $k$  is meant as a fixed index.

Inserting everything into 1 and the fact that all the terms remain unchanged except  $z_k$ , which is set to 1, we get the result

$$p(z'_k = 1 \mid z_{\setminus k}) = \frac{1}{1 + \exp \left( -2 \sum_j z'_k W_{kj} z'_j - b_k \right)} = \frac{1}{1 + \exp \left( -2 \sum_j W_{kj} z_j - b_k \right)} = \sigma(u_k)$$

One could do the same calculation for  $z_k^{n+1} = 0$  and would of course find that

$$p(z'_k = 0 \mid z_{\setminus k}) = 1 - \sigma(u_k)$$

(b)

The equality is trivially verified if  $z_k = z'_k$ , hence we assume  $z_k \neq z'_k$ . We have then to prove that

$$p_T(z_k^{n+1} = 0 \mid z_k^n = 1, \mathbf{z}_{\setminus k}) p(z_k^n = 1 \mid \mathbf{z}_{\setminus k}) = p_T(z_k^{n+1} = 1 \mid z_k^n = 0, \mathbf{z}_{\setminus k}) p(z_k^n = 0 \mid \mathbf{z}_{\setminus k})$$

or

$$\frac{p_T(z_k^{n+1} = 0 \mid z_k^n = 1, \mathbf{z}_{\setminus k})}{p_T(z_k^{n+1} = 1 \mid z_k^n = 0, \mathbf{z}_{\setminus k})} = \frac{p(z_k^n = 0 \mid \mathbf{z}_{\setminus k})}{p(z_k^n = 1 \mid \mathbf{z}_{\setminus k})} \quad (4)$$

Since the updating probability of the term  $k$  does not depend on the value of the latter at the previous step, but just relies on the value of all the other components and the new value for  $z_k$  we may write

$$p_{\text{T}}(z_k^{n+1} \mid z_k^n, \mathbf{z}_{\setminus k}) = p_{\text{T}}(z_k^{n+1} \mid \mathbf{z}_{\setminus k}) = \begin{cases} \sigma(u_k) & \text{if } z_k^{n+1} = 1 \\ 1 - \sigma(u_k) & \text{if } z_k^{n+1} = 0 \end{cases}$$

The right hand-side member of equation 4 can be rewritten using 2 as

$$\frac{p(z_k^n = 0 \mid \mathbf{z}_{\setminus k})}{p(z_k^n = 1 \mid \mathbf{z}_{\setminus k})} = \frac{p(z_k^n = 0, \mathbf{z}_{\setminus k})}{p(z_k^n = 1, \mathbf{z}_{\setminus k})} = \frac{p(\mathbf{z})}{p(\mathbf{z}')}$$

and equality 4 follows from 3.

# BIC #7

## EX 03

$$E(z) = \frac{1}{2} z^t W z + b^t z$$

Top :  $z_k \in \{-1, 1\}$

Bottom :  $z_k \in \{0, 1\}$

a)  $W_{ks} = 4 \hat{W}_{ks}$   
 $b_k = 2 \hat{b}_k - 2 \sum_{s=1}^k \hat{W}_{ks}$

Map  $\Omega = \{0, 1\}^k \rightarrow \hat{\Omega} = \{-1, 1\}^k$

$$E(z) = E(\hat{z}) + C \quad \forall \hat{z} = 2z - 1 \text{ and } C = \frac{1}{2} \sum_{k,s=1}^k \hat{W}_{ks} - \sum_{k=1}^k \hat{b}_k$$

$$E(z) = 2 z^t \hat{W} z + 2 \hat{b}^t z - 2 \sum_{s=1}^k \hat{W}_{sk} z_k \Rightarrow \text{here implicit } \sum_k$$

$$E(\hat{z} = 2z - 1) = 2z^t \hat{W} z + \frac{1^t \hat{W} 1}{2} + 2 \hat{b}^t z - b^t 1 \quad \text{where } 1_k = 1$$

$$\underbrace{- \frac{1^t \hat{W} z}{2} - z^t \frac{\hat{W} 1}{2}}_{\text{sum over } k} = -2 \sum_{s=1}^k \hat{W}_{sk} z_k$$

then,  $-b^t 1 = -\sum_k b_k$  and  $\frac{1^t \hat{W} 1}{2} = \frac{1}{2} \sum_{k,s=1}^k \hat{W}_{ks}$

b)  $p(z) = \frac{1}{z} e^{-E(z)} \rightarrow p'(z) = \frac{1}{z'} e^{-E(z) + C}$

$$p'(z) = \frac{1}{z'} e^C e^{-E(z)} \Rightarrow \frac{1}{z'} = \frac{e^C}{z'}$$

the normalized factor  
reestablish the  
correct probability

$\rightarrow$  the probability does not change

$$[\text{Also can be seen as } z' = \sum_k e^C e^{-E(z)} = e^C \sum_k e^{-E(z)} = e^C z]$$

c)  $b_1 = b_2 = 0 \quad W_{12} = W_{21} \geq 0$

$$\Omega: \begin{pmatrix} 00 \\ 01 \\ 10 \\ 11 \end{pmatrix} \begin{matrix} z^1 \\ z^2 \\ z^3 \\ z^4 \end{matrix} \quad \hat{\Omega}: \begin{pmatrix} -1 & -1 \\ -1 & 1 \\ 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{matrix} \hat{z}_1^1 \\ \hat{z}_1^2 \\ \hat{z}_2^3 \\ \hat{z}_2^4 \end{matrix}$$

$$W_{12} = 0 : E(z) = 0 \quad \forall z \in \Omega, E(\hat{z}) = 0 \quad \forall \hat{z} \in \hat{\Omega}$$

If we increase  $W_{12}$ , we obtain always a positive contribution (or null) from  $z^t W z$  in  $\Omega$ .

In  $\hat{\Omega}$ ,  $2 z^t W z$  counts an extra contribution, that is balanced from the term  $-2 \sum_{s=1}^k W_{ks} z_k$ . Part of  $W$  act then as a bias term in  $\hat{\Omega}$ .