

# Grundlagen der Bioinformatik

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1	2	3	4	$\Sigma$

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## Blatt 9

(Abgabe am 07.06.2022)

### Theoretical Assignments

#### Task 1: Viterbi Algorithm by hand (2)

V	b	R	Y	Y	e
b	1	0	0	0	0.005762
P	0	0.28	0.0378	0.057624	0
N	0	0.09	0.1372	0.019208	0

$$\begin{aligned}V_N(2) &= e_N(Y) * \max(v_k(1) * p_{KN}) \\&= 0.7 * \max(0.09 * 0.2, 0.28 * 0.7) \\&= 0.7 * \max(0.018, 0.196) \\&= 0.7 * 0.196 \\&= 0.1372\end{aligned}$$

$$\begin{aligned}V_P(3) &= e_P(Y) * \max(v_k(2) * p_{KP}) \\&= 0.6 * \max(0.00378 * 0.2, 0.1372 * 0.7) \\&= 0.6 * \max(0.000756, 0.09604) \\&= 0.6 * 0.09604 \\&= 0.057624\end{aligned}$$

$$\begin{aligned}V_N(3) &= e_N(Y) * \max(v_k(2) * p_{KN}) \\&= 0.7 * \max(0.1372 * 0.2, 0.00378 * 0.7) \\&= 0.7 * \max(0.02744, 0.002646) \\&= 0.7 * 0.02744 \\&= 0.019208\end{aligned}$$

Termination:

$$\begin{aligned}
 P(x, \pi^*) &= \max_{k \in Q} (v_k(L) p_{ke}) = \max(0.057624 * 0.1, 0.019208 * 0.1) \\
 &= \max(0.0057624, 0.0019208) \\
 &= 0.0057624
 \end{aligned}$$

Traceback:

V	b	R	Y	Y	e
<b>b</b>	<b>1</b>	0	0	0	<b>0.005762</b>
<b>P</b>	0	<b>0.28</b>	0.0378	<b>0.057624</b>	0
<b>N</b>	0	0.09	<b>0.1372</b>	0.019208	0

$$\begin{aligned}
 \pi_3^* &= \operatorname{argmax}_{k \in Q} (v_k(L) p_{ke}) = Y_P \\
 \pi_2^* &= \operatorname{ptr}_3(\pi_3^*) = \operatorname{ptr}_3(Y_P) = Y_N \\
 \pi_1^* &= \operatorname{ptr}_2(\pi_2^*) = \operatorname{ptr}_2(Y_N) = R_P \\
 \pi_1^* &= \operatorname{ptr}_1(\pi_1^*) = \operatorname{ptr}_1(R_P) = b \\
 \implies \pi^* &= b R_P Y_N Y_P \text{ Final comparison:} \\
 \mathbb{P}(R Y Y, R_P, Y_N, Y_P) &= 0.28 * (0.7 * 0.7 * 0.6 * 0.7 * 0.01) = 0.0057624
 \end{aligned}$$

## Task 2: Profile HMMs (6)

$$\begin{aligned}
 e_{m1}(A) &= \frac{5}{10} & e_{m1}(T) &= \frac{5}{10} & e_{m1}(X) &= 0 & \text{with } X \neq \{A, T\} \\
 e_{m2}(A) &= \frac{1}{10} & e_{m2}(C) &= \frac{4}{10} & e_{m2}(G) &= \frac{4}{10} & e_{m2}(X) = \frac{1}{10} \text{ with } X \neq \{A, C, G\} \\
 e_{m3}(T) &= \frac{2}{10} & e_{m3}(C) &= \frac{2}{10} & e_{m3}(X) &= \frac{6}{10} & \text{with } X \neq \{C, T\} \\
 e_{m4}(C) &= \frac{4}{10} & e_{m4}(G) &= \frac{4}{10} & e_{m4}(T) &= \frac{2}{10} & e_{m4}(X) = 0 \text{ with } X \neq \{C, G, T\} \\
 e_{m5}(C) &= \frac{2}{10} & e_{m5}(G) &= \frac{6}{10} & e_{m5}(X) &= \frac{2}{10} & \text{with } X \neq \{C, G\}
 \end{aligned}$$

The third column refers to an insert state because more than 50% are gaps.

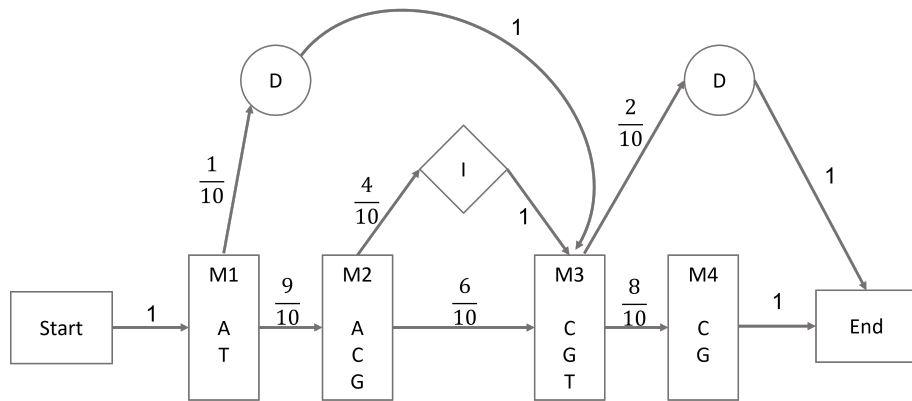


Figure 1: Our state diagram

### Task 3: Supervised training (2)

$$p_{kl} = \frac{P_{kl} + 1}{\sum_{q \in Q} P_{kq} + 1}$$

$$e_k(b) = \frac{E_k(b) + 1}{\sum_{s \in \Sigma} E_k(s) + 1}$$

## Practical Assignments

### Task 4: Decoding data using the Viterbi algorithm (10)

For Task 4 we used Python 3.8.8, and the following libraries **io**, **pandas**, **sys**, **getopt**, **Bio**, **collections** and **numpy**.

Enter the following code in the command line to run the file:

```
python auckenthaler_dittschar_hmm_handler.py -a cpg.hmm -s input_hmm.fasta
```

We exported the output as a txt-file and saved each Sequence in a Seperate txt-file named:

- auckenthaler\_dittschar\_viterbi\_Seq1.txt
- auckenthaler\_dittschar\_viterbi\_Seq2.txt
- auckenthaler\_dittschar\_viterbi\_Seq3.txt
- auckenthaler\_dittschar\_viterbi\_Seq4.txt