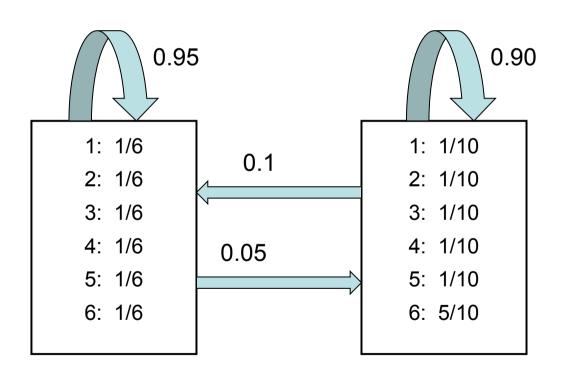


Royal Institute of Technology

## DGM APPLICATION HMM

# EMISSION & TRANSITION DISTRIBUTIONS



$$p(\boldsymbol{x}_{1:T}, \boldsymbol{z}_{1:T+1})$$

$$= p(\boldsymbol{x}_{1:T}|\boldsymbol{z}_{1:T})p(\boldsymbol{z}_{1:T+1})$$

$$= p(\boldsymbol{z}_1) \left(\prod_{t=1}^T p(\boldsymbol{z}_{t+1}|\boldsymbol{z}_t)\right) \left(\prod_{t=1}^T p(\boldsymbol{x}_t|\boldsymbol{z}_t)\right)$$

Categorial (or Gaussian)

- Starts in the state  $z_1=k^*$
- When in state z<sub>t</sub>
  - emits  $p(x_t|z_t)$

transits to p(z<sub>t+1</sub>|z<sub>t</sub>)  $lacktrlant{A}_{z_{t+1},z_t}$ 

- Stops when reaching a stop state
- $\star$  z<sub>T+1</sub> is assumed to be stop

The parameters Now given!!

# AN HMM CAN BE SEEN AS A DGM

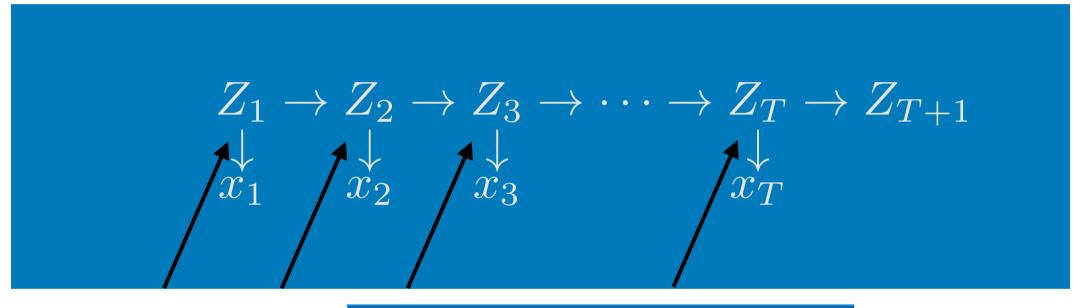
- Z<sub>i</sub> hidden
- Xi observable
- Hidden often not observable when training, never when applying

All the same

State	1	2	3	4
1	A <sub>11</sub>	$A_{21}$	A <sub>31</sub>	A <sub>41</sub>
2	A <sub>12</sub>	$A_{22}$	A <sub>32</sub>	A <sub>42</sub>
3	$A_{13}$	$A_{23}$	$A_{33}$	$A_{43}$
4				A <sub>44</sub>

## FOR 4 STATES HMM

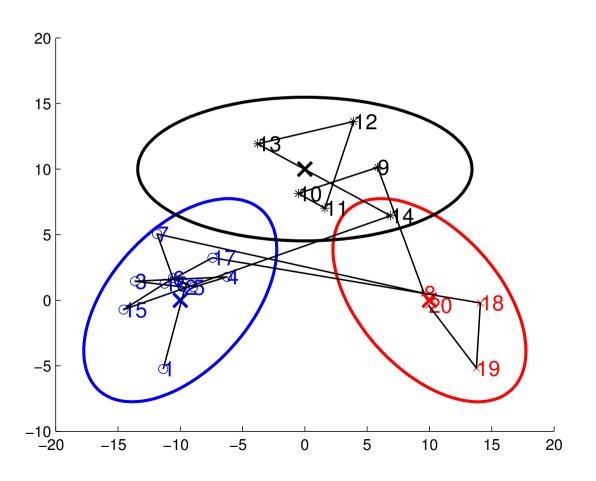
#### EMISSION PROBABILITIES - HMM WITH 4 STATES & 3 SYMBOLS

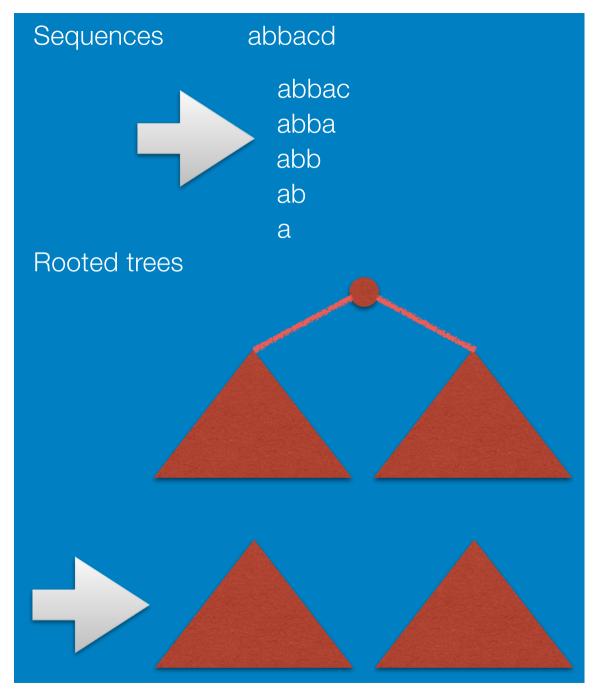


All the same

State\Symb	1	2	3
1	B <sub>11</sub>	B <sub>21</sub>	B <sub>31</sub>
2	B <sub>12</sub>	B <sub>22</sub>	B <sub>32</sub>
3	B <sub>13</sub>	B <sub>23</sub>	B <sub>33</sub>
4	B <sub>14</sub>	B <sub>24</sub>	B <sub>34</sub>

### GAUSSIAN EMISSIONS AND HIDDEN STATES





- What is a subproblem?
- What is a subsolution?
- How do we decompose into smaller subproblems?
- How do we combine subsolutions into larger?
- How do we enumerate?
- How many and what time?

Polynomial many

Polynomial time

Polynomial time

Polynomial time

Polynomial time overall

What is a subproblem?

What is a subsolution?

How do we decompose into smaller subproblems?

How do we combine subsolutions into larger?

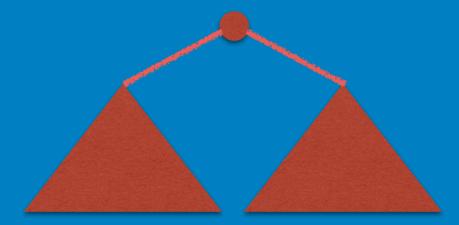
How do we enumerate?

How many and what time?

#### Sequences

$$\begin{array}{ccc} Z_1 \rightarrow Z_2 \rightarrow Z_3 \\ \downarrow & \downarrow & \downarrow \\ x_1 & x_2 & x_3 \end{array}$$

#### Rooted trees



$$P(A) = \sum_{i} P(A, B = i) = \sum_{i} P(A | B = i)P(B = i)$$

#### DP, SUM RULE, & CONDITIONING

 How do we decompose into smaller subproblems?

Dynamic programming is similar to divide + conquet in that it solves a problem by dividing it into sub-problems. However, in the dynamic programming paradigm, the larger problem is solved by solving and remembering overlapping sub-problems, which are reased repeatedly in the process.



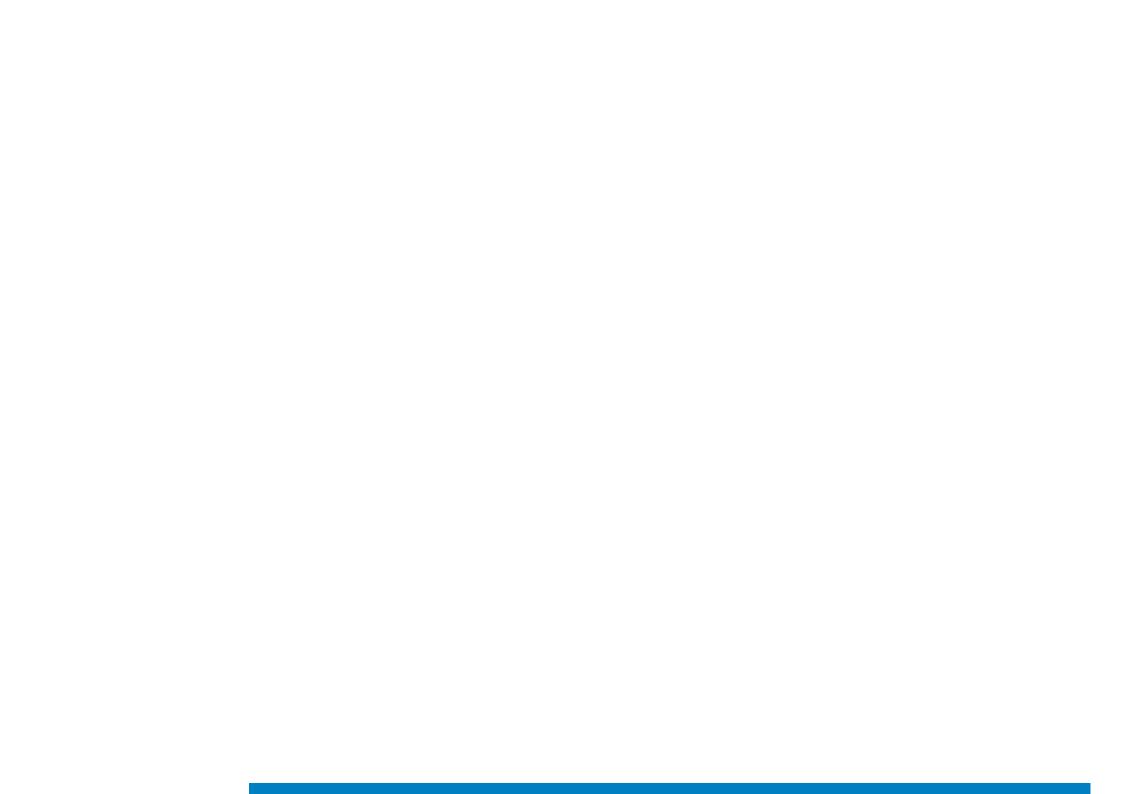
## Applying sum rule

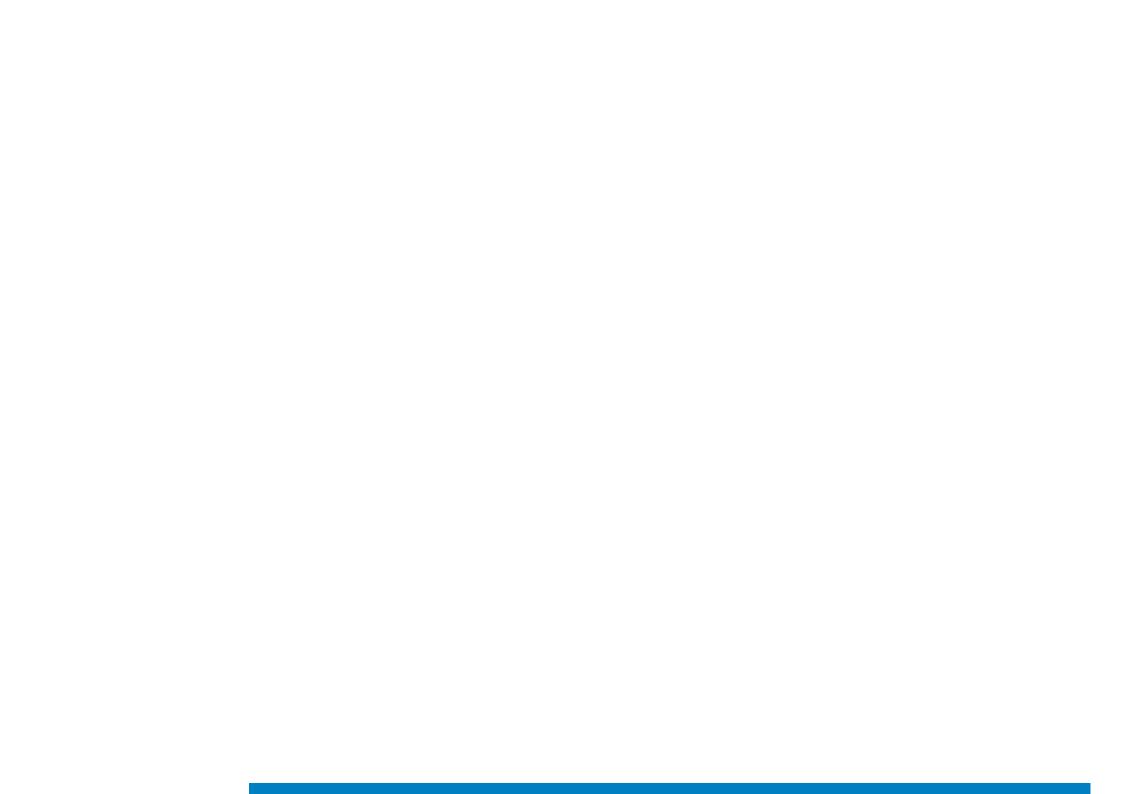
Notice, by the sum rule,

$$f_t(k) = p(x_{1:t-1}, Z_t = k) = \sum_{k' \in [K]} p(x_{1:t-1}, Z_{t-1} = k', Z_t = k)$$
 The set of states

each term in the sum is a probability of an event

which, as noted, can be broken into smaller





## Forward recursion

$$f_t(k) = \sum_{k'} \underbrace{f_{t-1}(k')}_{\text{smaller}} \underbrace{p(\boldsymbol{x}_{t-1}|\boldsymbol{Z}_{t-1}=k')}_{\text{emission}} \underbrace{p(\boldsymbol{Z}_t=k|\boldsymbol{Z}_{t-1}=k')}_{\text{transition}}$$

## Backward variable

Defined by

$$b_t(k) := p(\boldsymbol{x}_{t+1:T}|\boldsymbol{Z}_t = k)$$

"Graphical model"

$$? \rightarrow Z_{t} = k \rightarrow ? \rightarrow \cdots ?$$

$$\downarrow \qquad \downarrow \qquad \downarrow \qquad \downarrow$$

$$? \qquad x_{t+1} \qquad x_{T}$$

## Backward recursion

$$b_t(k) := p(\boldsymbol{x}_{t+1:T}|\boldsymbol{Z}_t = k)$$

DP also for the backward variable bt

$$b_t(k) = \sum_{l} \underbrace{p(\boldsymbol{Z}_{t+1} = l | \boldsymbol{Z}_t = k)}_{\text{transition}} \underbrace{b_{t+1}(l)}_{\text{"smaller"}} \underbrace{p(\boldsymbol{x}_{t+1} | \boldsymbol{Z}_{t+1} = l)}_{\text{emission}}$$