

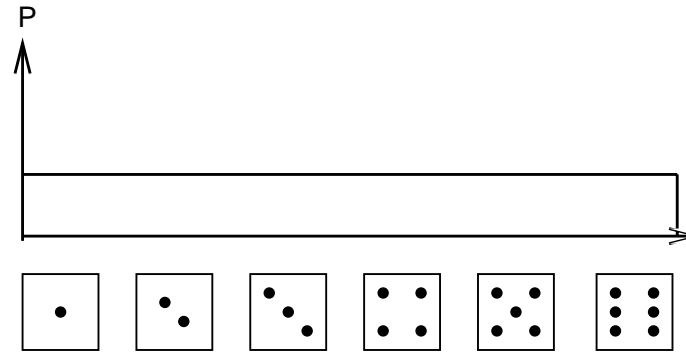
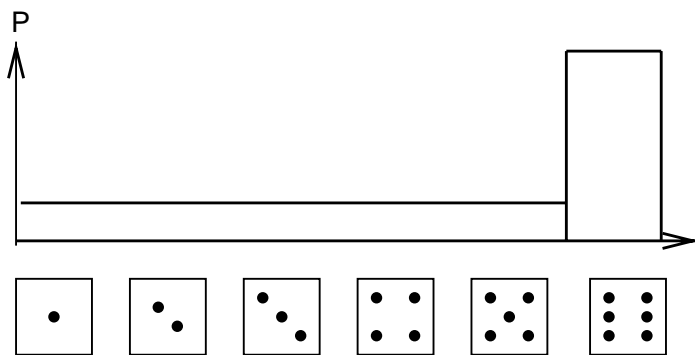
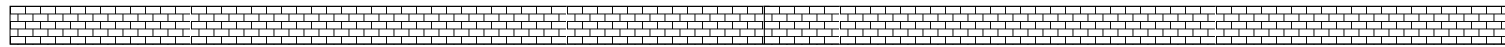
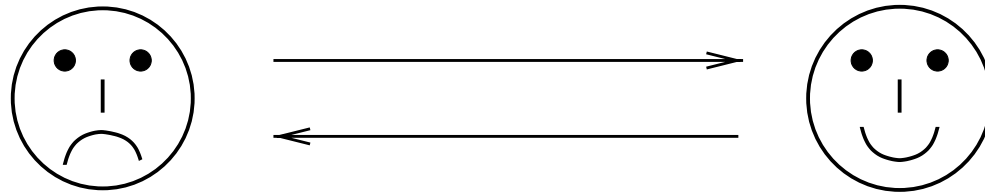
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# A Tutorial on Hidden Markov Models

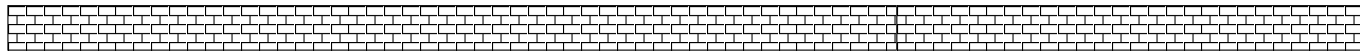
Dirk Husmeier  
Biomathematics and Statistics Scotland  
at the Scottish Crop Research Institute  
Invergowrie, Dundee DD2 5DA, UK  
Email: [dirk@bioss.ac.uk](mailto:dirk@bioss.ac.uk)  
<http://www.bioss.ac.uk/~dirk>

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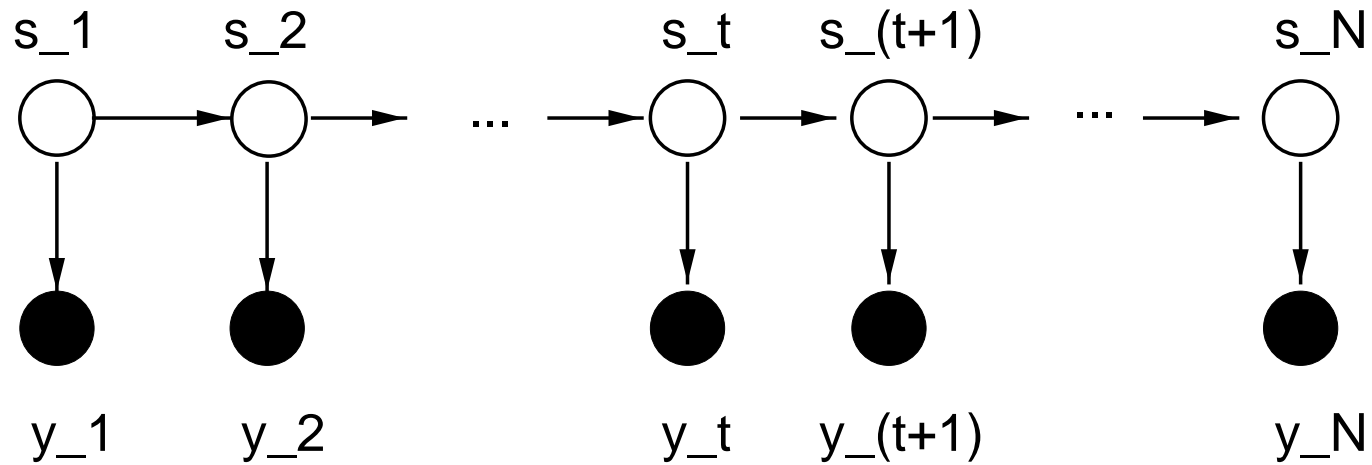
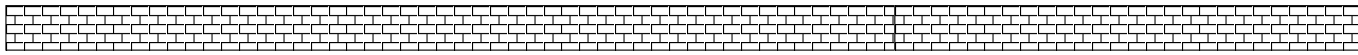
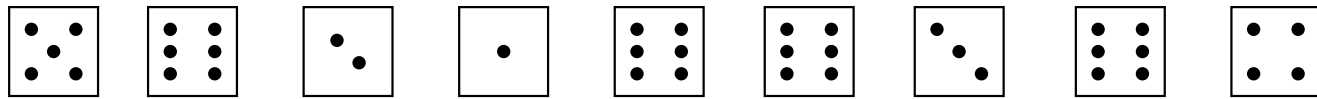
## Example: The Occasionally Corrupt Casino



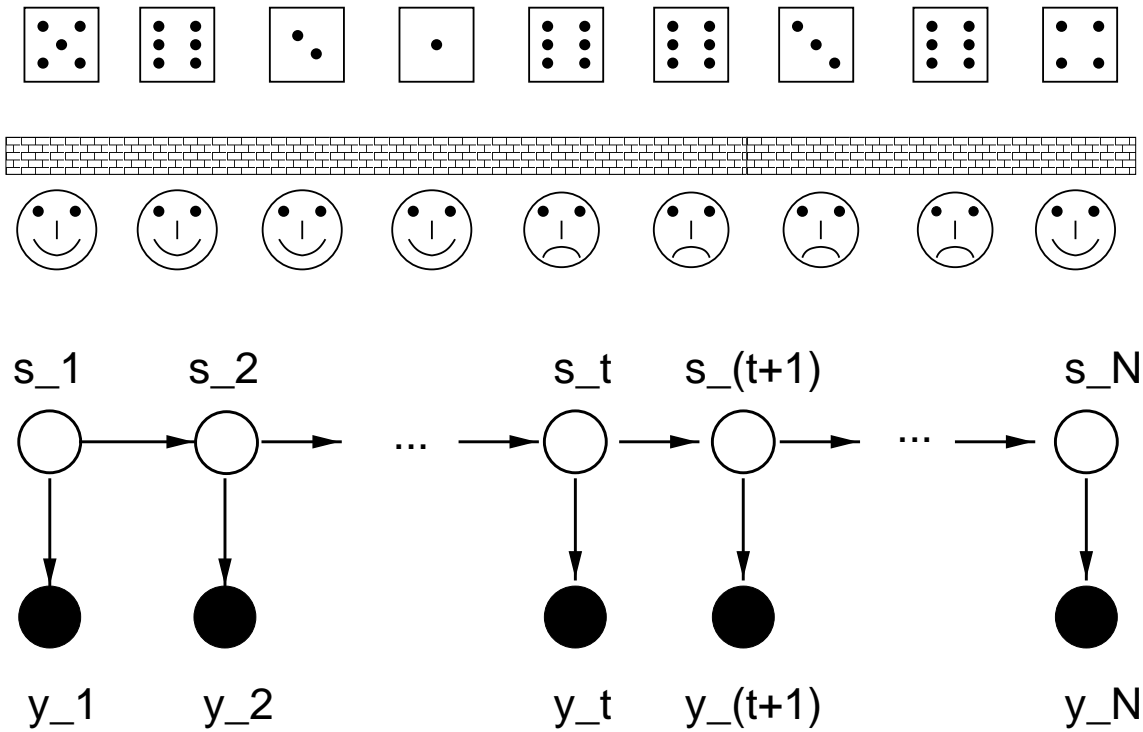
# Example: HMM



## Example: HMM



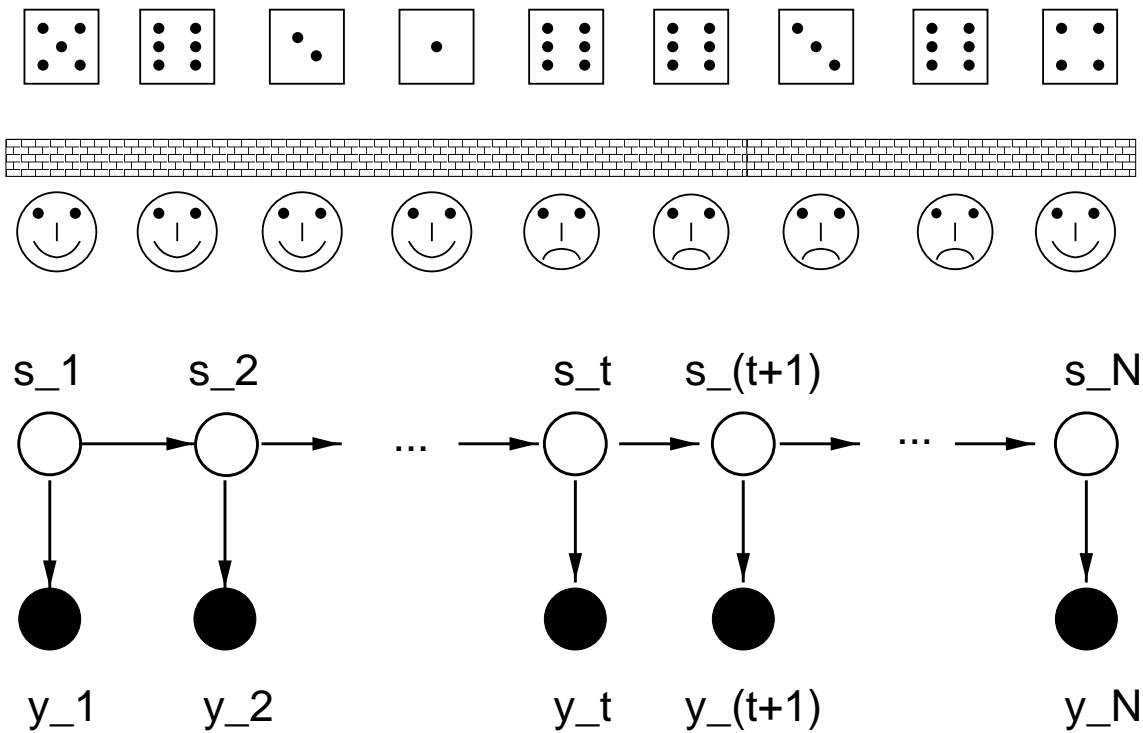
## The Most Likely State Sequence



Find the mode of  $P(S_1, \dots, S_N | y_1, \dots, y_N)$

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## The Most Likely State Sequence

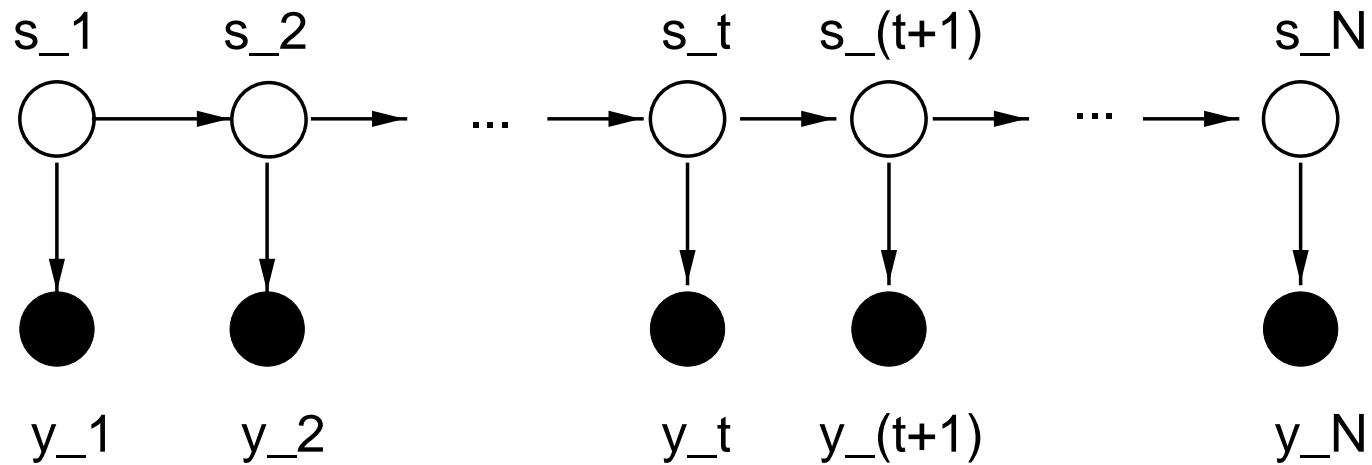


Find the mode of  $P(S_1, \dots, S_N | y_1, \dots, y_N)$

**Problem:**  $(S_1, \dots, S_N) : 2^N$  different sequences.

## Factorisation in HMMs

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$$\begin{aligned} P(y_1, \dots, y_N, S_1, \dots, S_N) &= \prod_{t=1}^N P(y_t | S_t) \prod_{t=2}^N P(S_t | S_{t-1}) P(S_1) \\ &:= \prod_{t=1}^N P(y_t | S_t) \prod_{t=1}^N P(S_t | S_{t-1}) \end{aligned}$$

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## The Most Likely State Sequence

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$$\begin{aligned}\max_{S_1, \dots, S_N} P(S_1, \dots, S_N | y_1, \dots, y_N) &= \max_{S_1, \dots, S_N} P(S_1, \dots, S_N, y_1, \dots, y_N) \\ &= \max_{S_N} \gamma_N(S_N)\end{aligned}$$

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$$\gamma_n(S_n) = \max_{S_1, \dots, S_{n-1}} P(y_1, \dots, y_n, S_1, \dots, S_n)$$

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## The Most Likely State Sequence

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## The Most Likely State Sequence

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$$\begin{aligned}\max_{S_1, \dots, S_N} P(S_1, \dots, S_N | y_1, \dots, y_N) &= \max_{S_1, \dots, S_N} P(S_1, \dots, S_N, y_1, \dots, y_N) \\ &= \max_{S_N} \gamma_N(S_N)\end{aligned}$$

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## The Most Likely State Sequence

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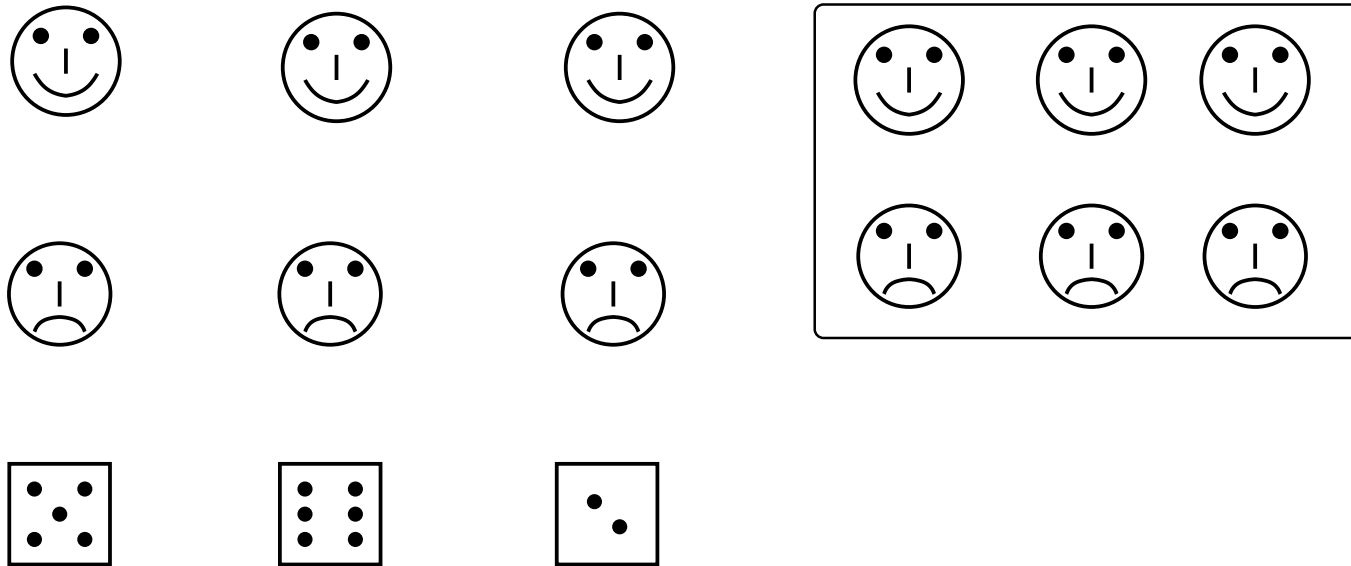
$$\begin{aligned}\max_{S_1, \dots, S_N} P(S_1, \dots, S_N | y_1, \dots, y_N) &= \max_{S_1, \dots, S_N} P(S_1, \dots, S_N, y_1, \dots, y_N) \\ &= \max_{S_N} \gamma_N(S_N)\end{aligned}$$

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Computational complexity  $K^N \rightarrow NK^2$  (K= number of states).

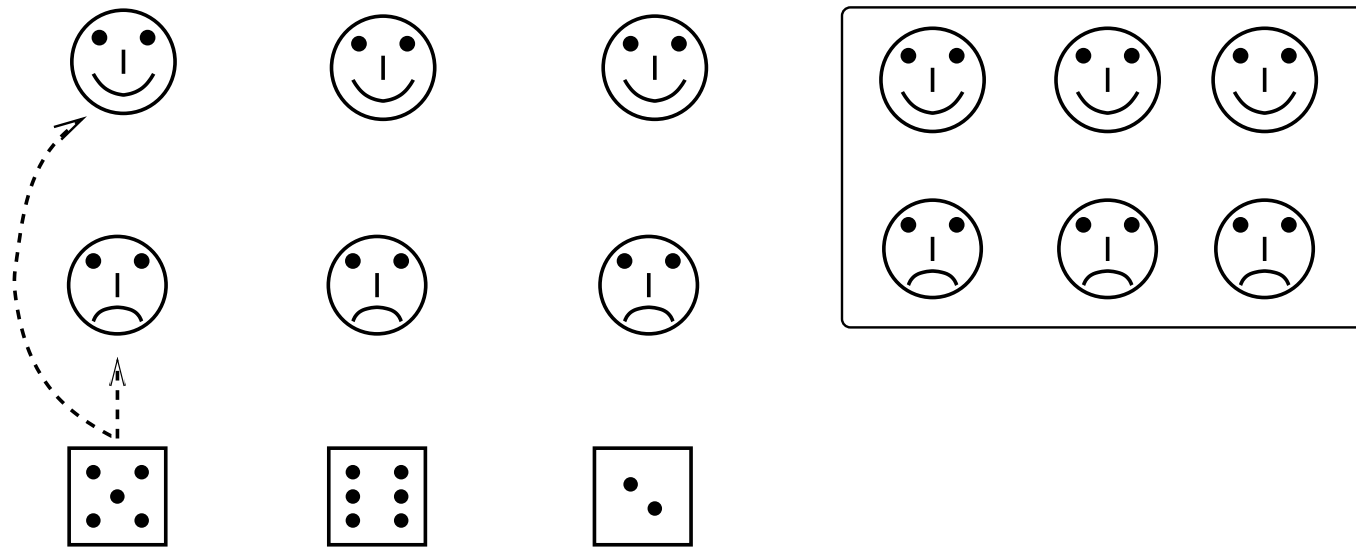
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## Viterbi Algorithm: Demo



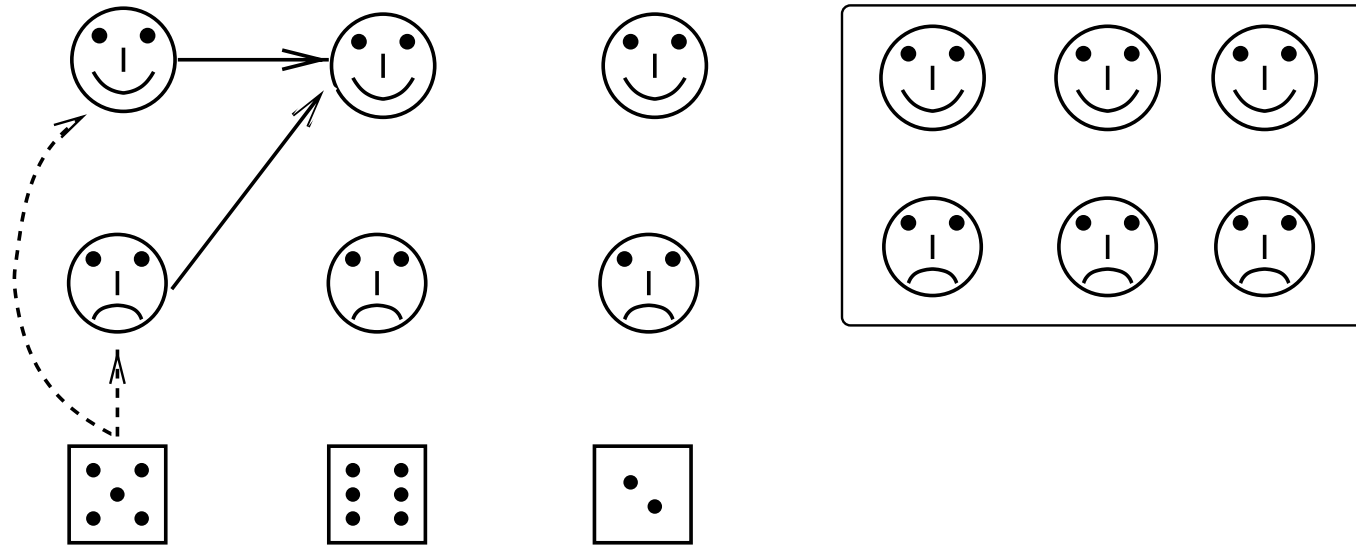
Initialisation	$\gamma_1(S_1) = P(y_1 S_1)P(S_1)$
Recursion ( $n = 2, \dots, N$ )	$\gamma_n(S_n) = P(y_n S_n) \max_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$ $\text{Pointer}(S_{n-1}) = \operatorname{argmax}_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$
Termination	$\hat{S}_N = \operatorname{argmax}_{S_N} \gamma_N(S_N)$
Traceback ( $n = N - 1, \dots, 1$ )	$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$

## Viterbi Algorithm: Demo



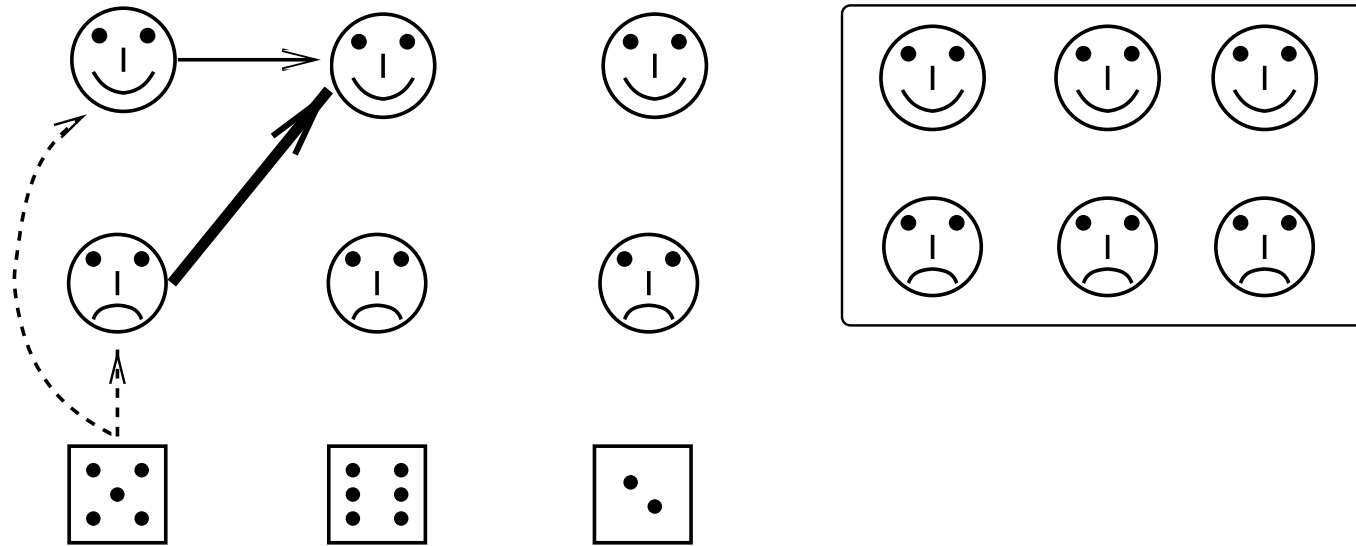
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## Viterbi Algorithm: Demo



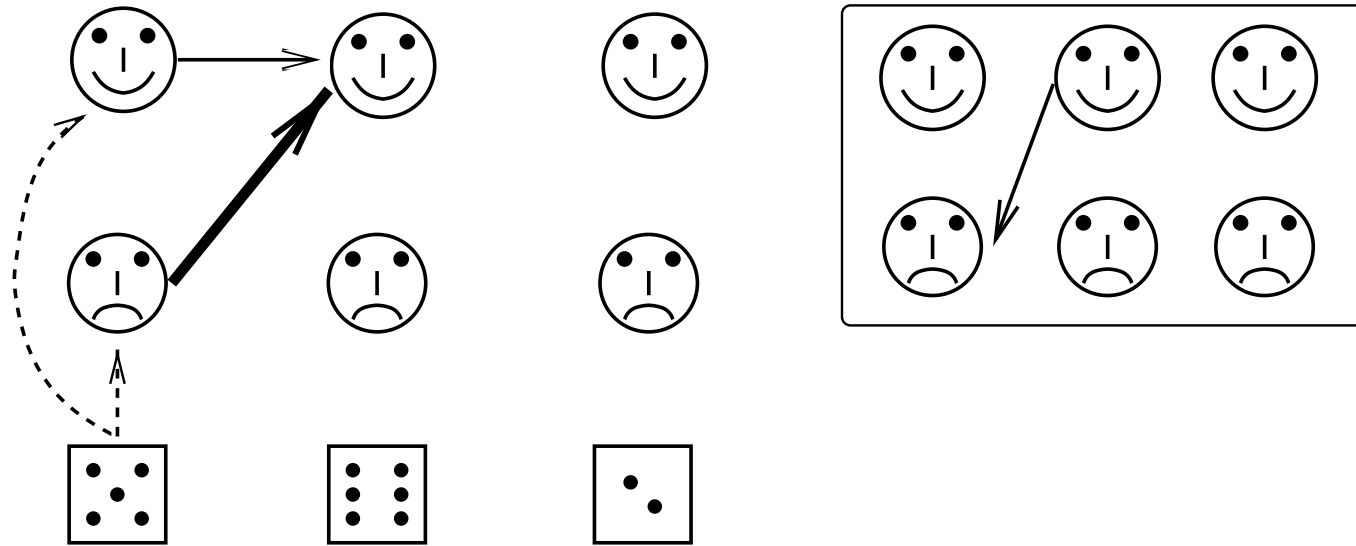
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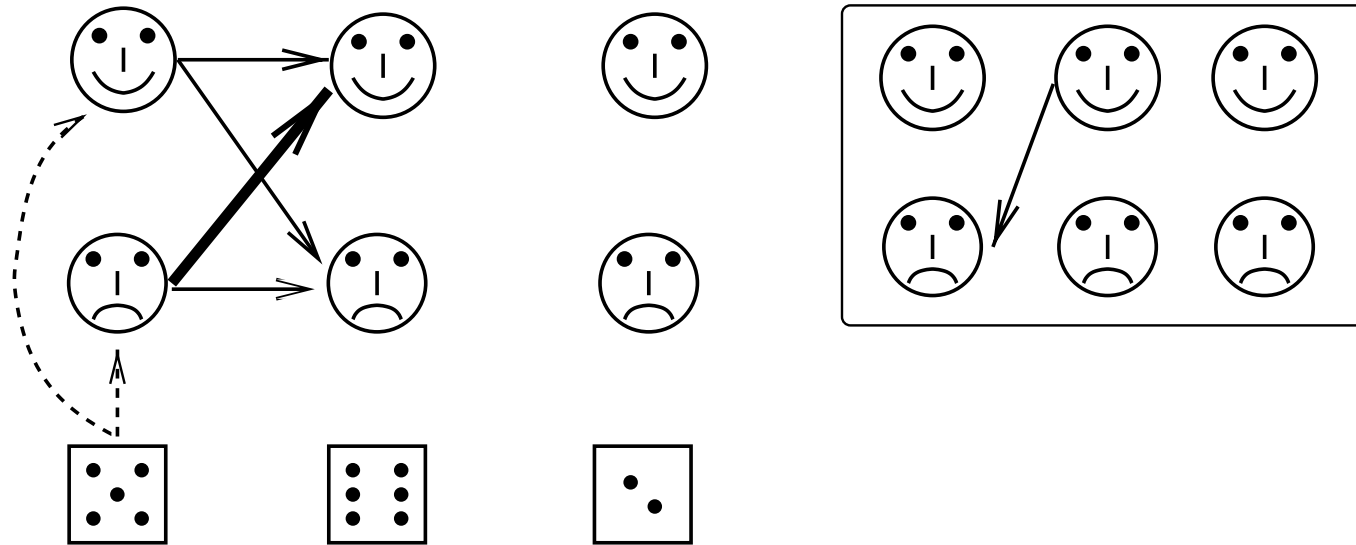
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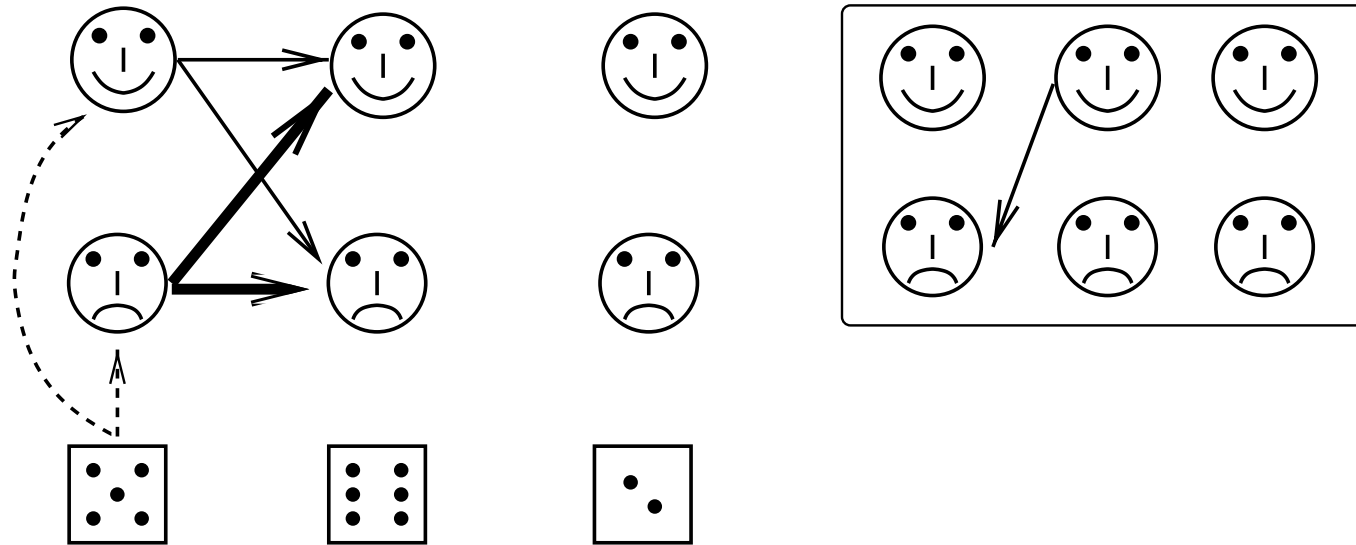
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## Viterbi Algorithm: Demo



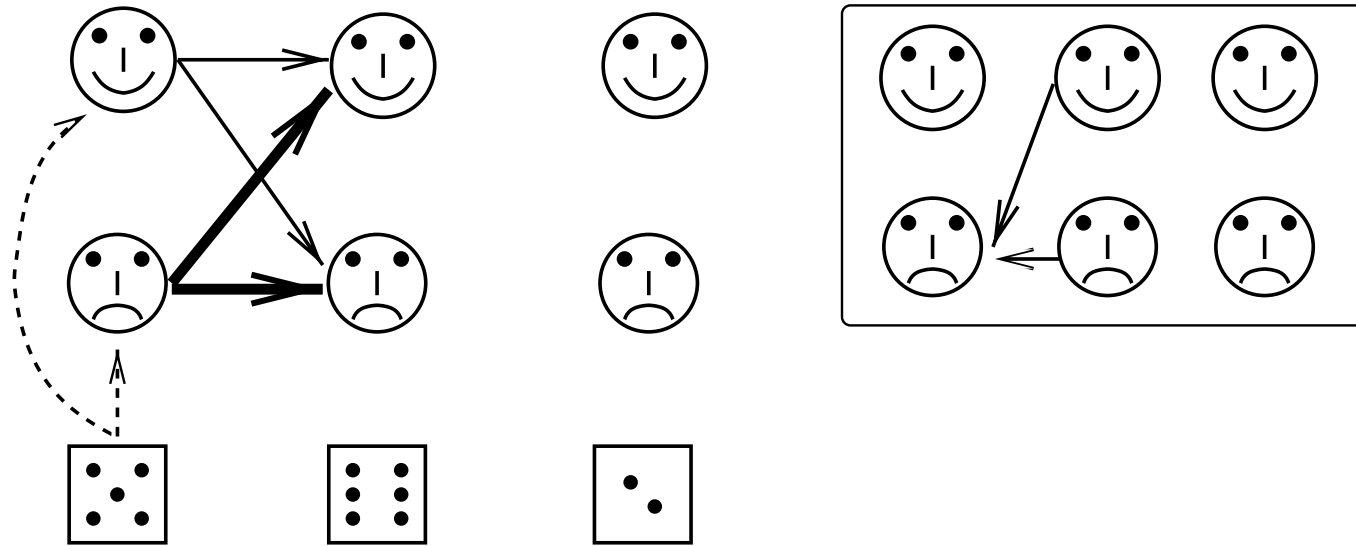
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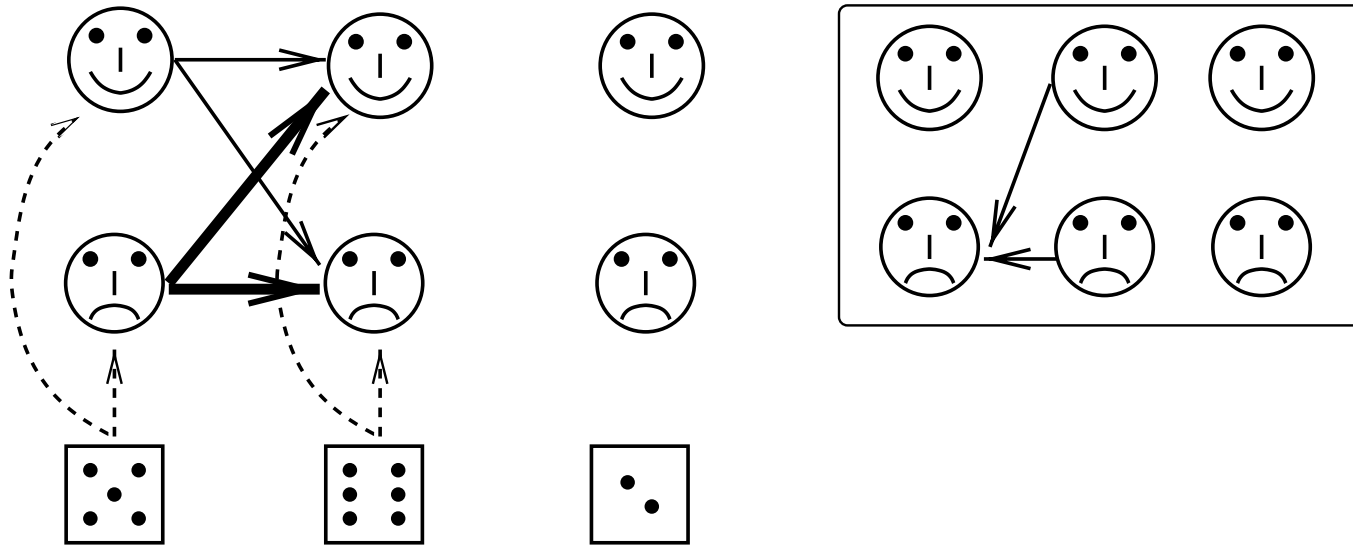
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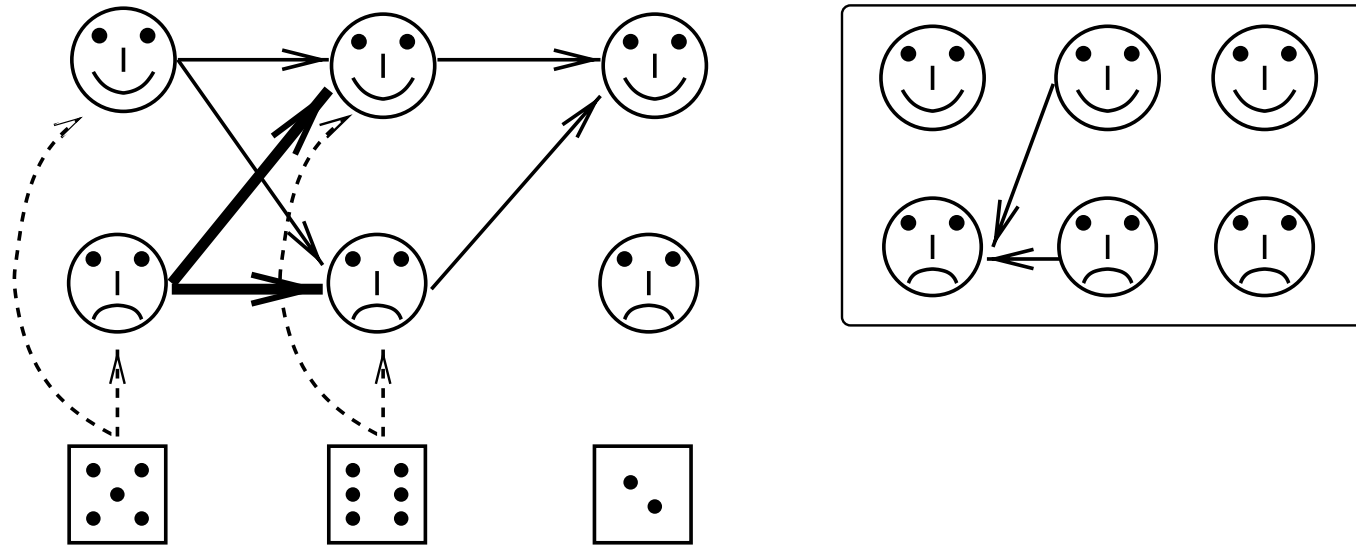
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## Viterbi Algorithm: Demo



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**Recursion** ( $n = 2, \dots, N$ )

$$\gamma_n(S_n) = P(y_n|S_n) \max_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

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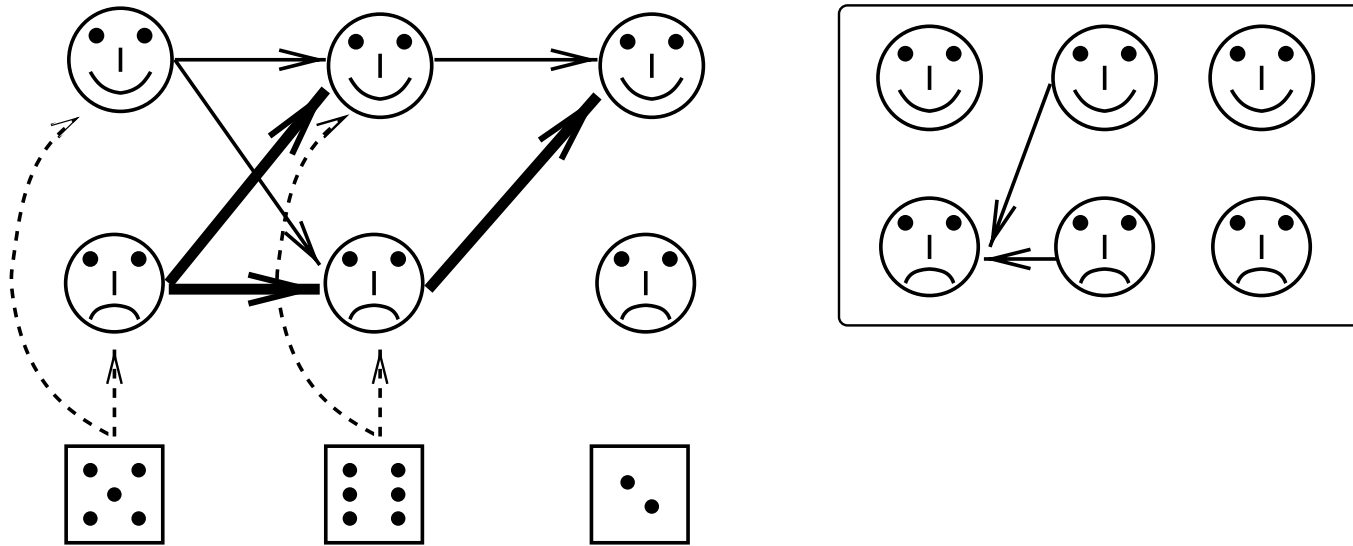
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Traceback ( $n = N - 1, \dots, 1$ )

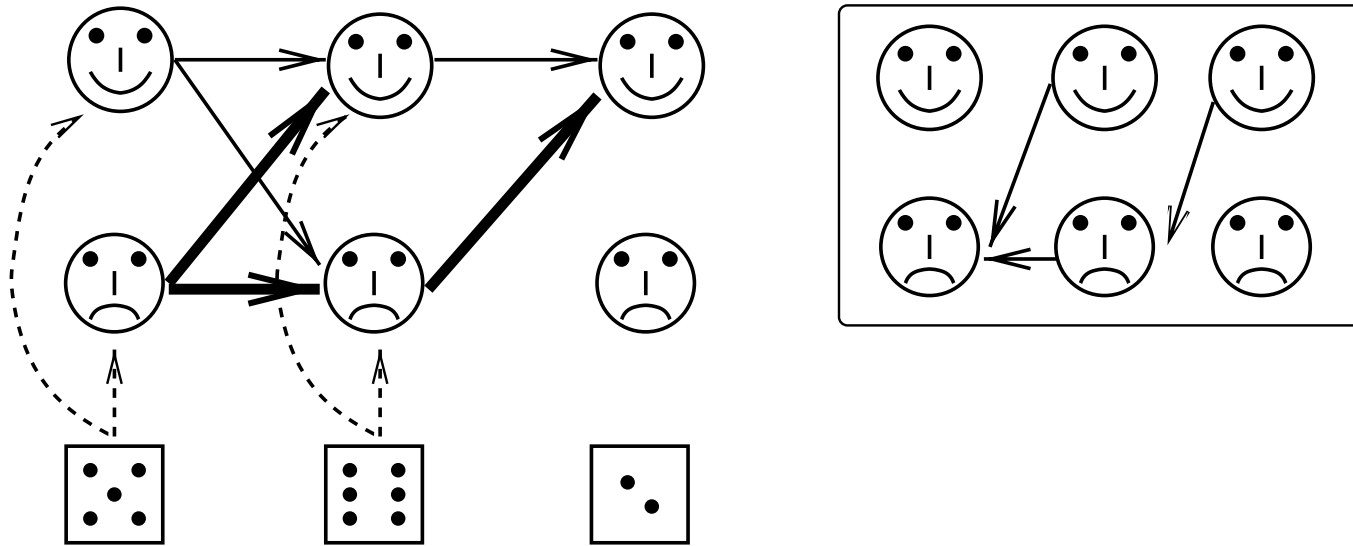
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## Viterbi Algorithm: Demo



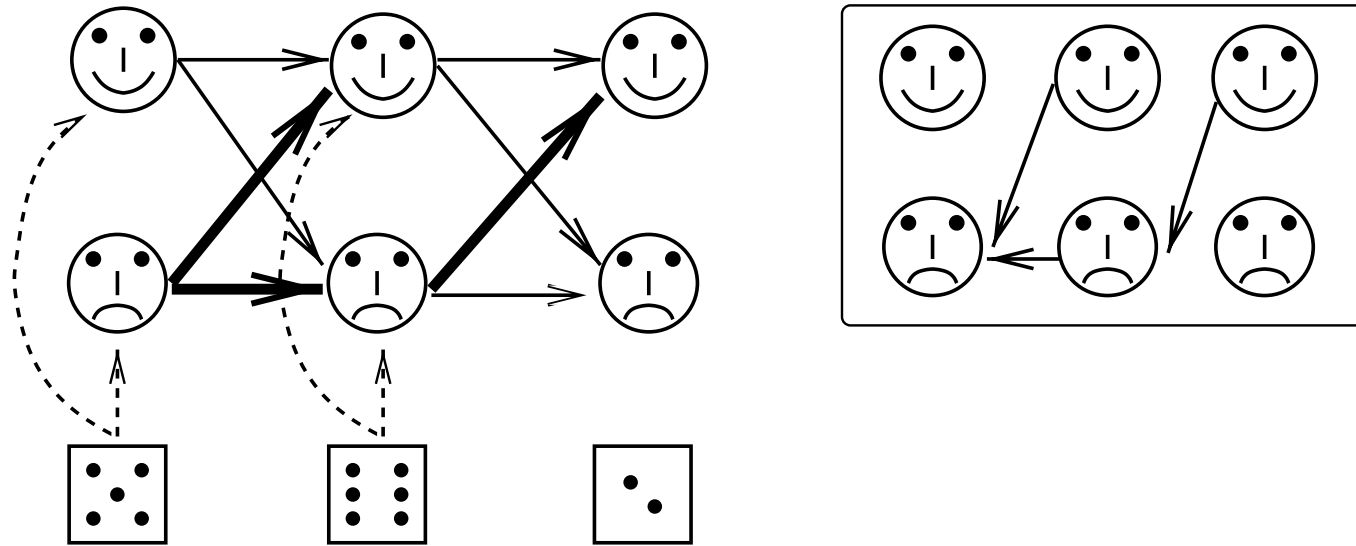
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## Viterbi Algorithm: Demo



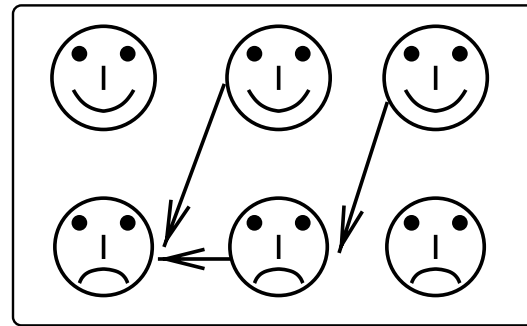
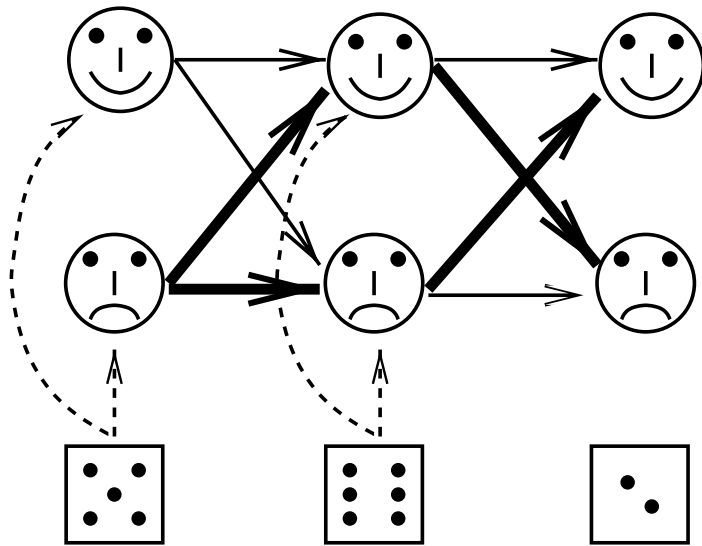
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## Viterbi Algorithm: Demo



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## Viterbi Algorithm: Demo



Initialisation

$$\gamma_1(S_1) = P(y_1|S_1)P(S_1)$$

**Recursion** ( $n = 2, \dots, N$ )

$$\gamma_n(S_n) = P(y_n|S_n) \max_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

$$\text{Pointer}(S_{n-1}) = \operatorname{argmax}_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

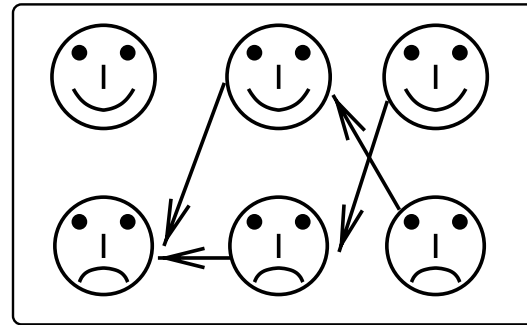
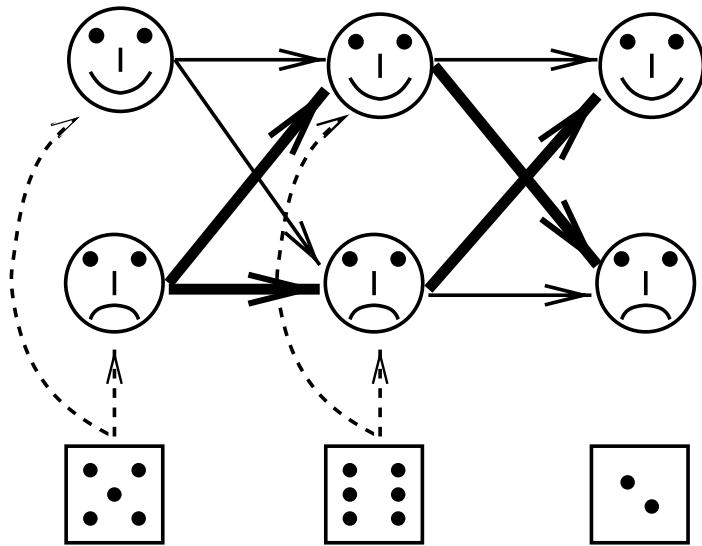
Termination

$$\hat{S}_N = \operatorname{argmax}_{S_N} \gamma_N(S_N)$$

Traceback ( $n = N - 1, \dots, 1$ )

$$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$$

## Viterbi Algorithm: Demo



Initialisation

$$\gamma_1(S_1) = P(y_1|S_1)P(S_1)$$

**Recursion** ( $n = 2, \dots, N$ )

$$\gamma_n(S_n) = P(y_n|S_n) \max_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

$$\text{Pointer}(S_n) = \operatorname{argmax}_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

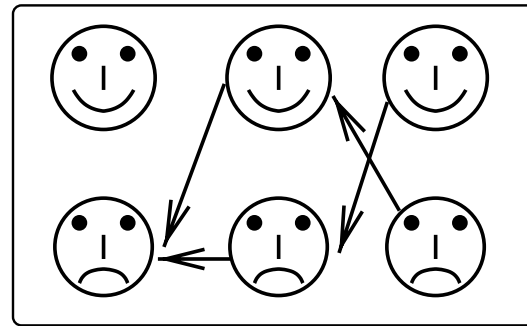
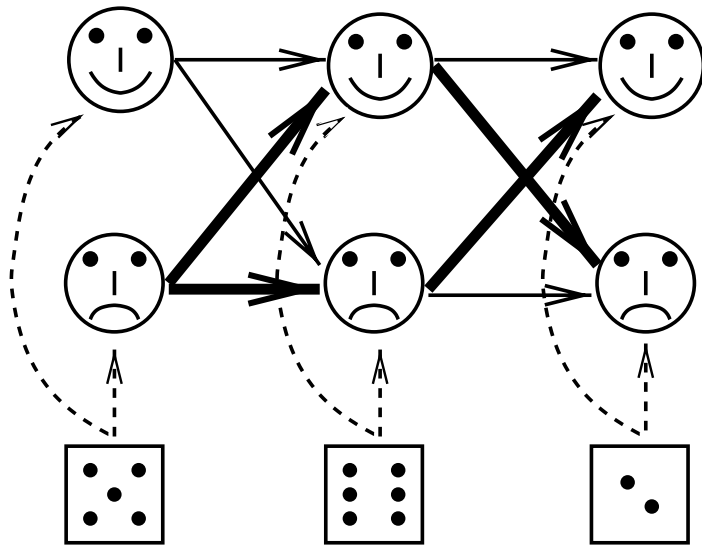
Termination

$$\hat{S}_N = \operatorname{argmax}_{S_N} \gamma_N(S_N)$$

Traceback ( $n = N - 1, \dots, 1$ )

$$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$$

## Viterbi Algorithm: Demo



Initialisation

$$\gamma_1(S_1) = P(y_1|S_1)P(S_1)$$

**Recursion** ( $n = 2, \dots, N$ )

$$\gamma_n(S_n) = P(y_n|S_n) \max_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

$$\text{Pointer}(S_{n-1}) = \operatorname{argmax}_{S_{n-1}} P(S_n|S_{n-1})\gamma_{n-1}(S_{n-1})$$

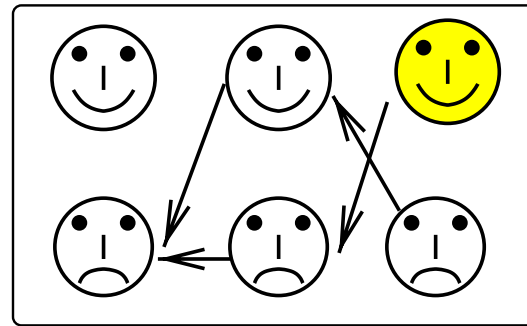
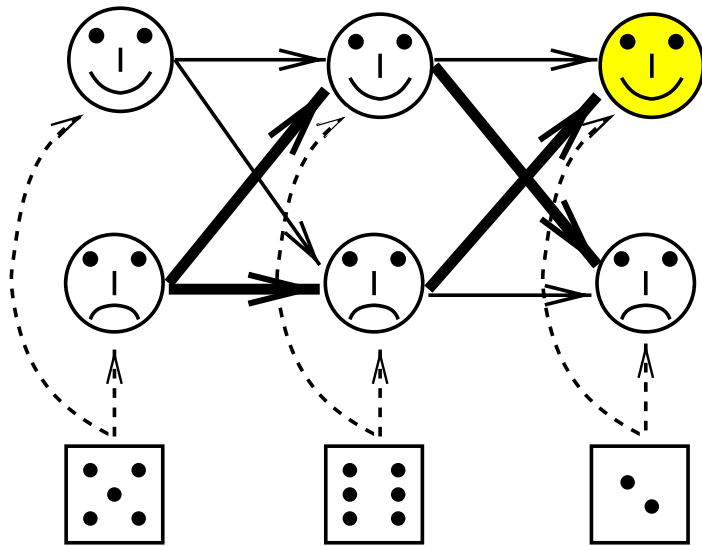
Termination

$$\hat{S}_N = \operatorname{argmax}_{S_N} \gamma_N(S_N)$$

Traceback ( $n = N - 1, \dots, 1$ )

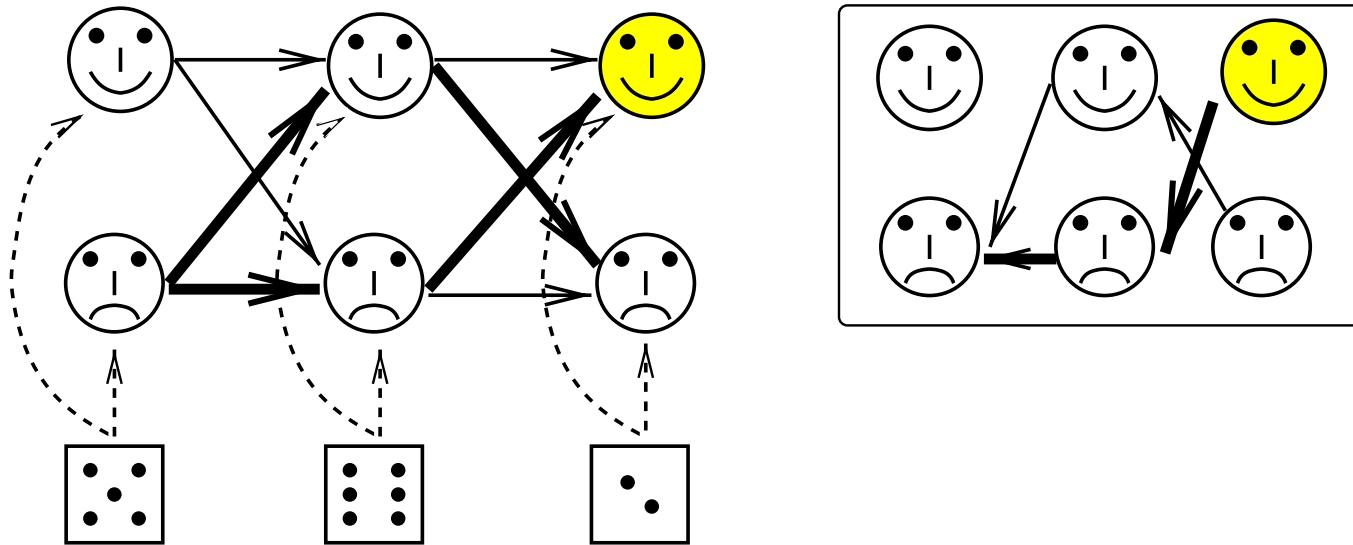
$$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$$

## Viterbi Algorithm: Demo



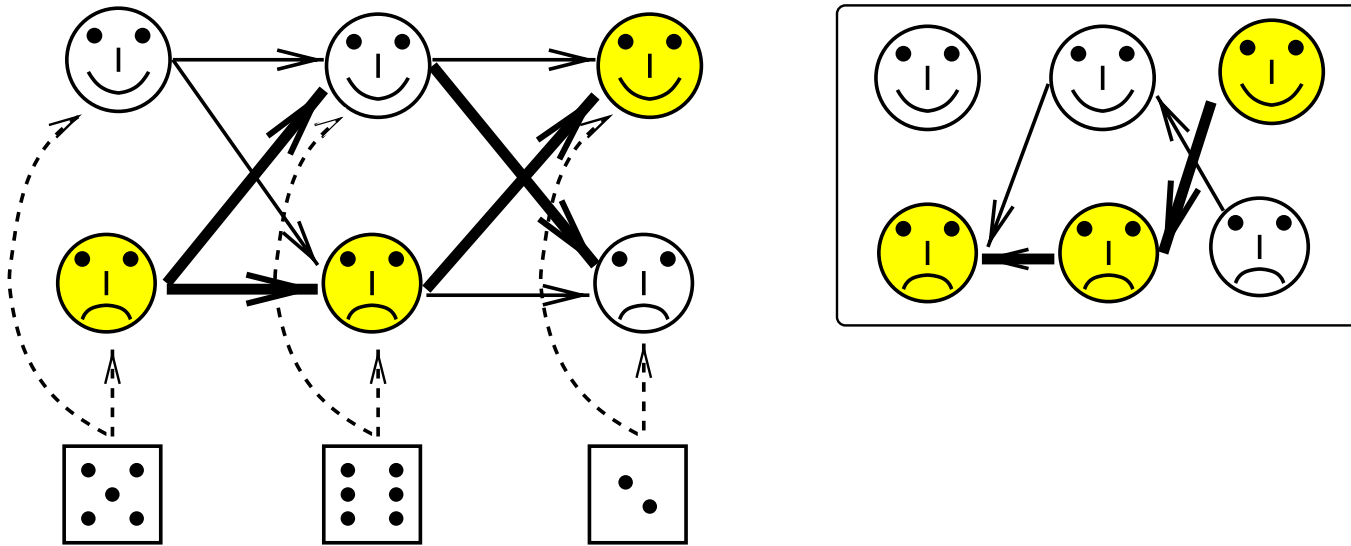
Initialisation	$\gamma_1(S_1) = P(y_1 S_1)P(S_1)$
Recursion ( $n = 2, \dots, N$ )	$\gamma_n(S_n) = P(y_n S_n) \max_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$ $\text{Pointer}(S_{n-1}) = \text{argmax}_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$
Termination	$\hat{S}_N = \text{argmax}_{S_N} \gamma_N(S_N)$
Traceback ( $n = N - 1, \dots, 1$ )	$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$

## Viterbi Algorithm: Demo



Initialisation	$\gamma_1(S_1) = P(y_1 S_1)P(S_1)$
Recursion ( $n = 2, \dots, N$ )	$\gamma_n(S_n) = P(y_n S_n) \max_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$ $\text{Pointer}(S_{n-1}) = \text{argmax}_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$
Termination	$\hat{S}_N = \text{argmax}_{S_N} \gamma_N(S_N)$
Traceback ( $n = N - 1, \dots, 1$ )	$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$

## Viterbi Algorithm: Demo



Initialisation	$\gamma_1(S_1) = P(y_1 S_1)P(S_1)$
Recursion ( $n = 2, \dots, N$ )	$\gamma_n(S_n) = P(y_n S_n) \max_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$ $\text{Pointer}(S_{n-1}) = \operatorname{argmax}_{S_{n-1}} P(S_n S_{n-1})\gamma_{n-1}(S_{n-1})$
Termination	$\hat{S}_N = \operatorname{argmax}_{S_N} \gamma_N(S_N)$
Traceback ( $n = N - 1, \dots, 1$ )	$\hat{S}_n = \text{Pointer}(\hat{S}_{n+1})$

## Logarithmic Version of the Viterbi Algorithm

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Prevent numerical underflow

$$\begin{aligned}\max_{S_1, \dots, S_N} P(S_1, \dots, S_N | y_1, \dots, y_N) &= \max_{S_1, \dots, S_N} \log P(S_1, \dots, S_N, y_1, \dots, y_N) \\ &= \max_{S_N} \log \gamma_N(S_N)\end{aligned}$$

$$\begin{aligned}\gamma_n(S_n) &= \max_{S_1, \dots, S_{n-1}} P(y_1, \dots, y_n, S_1, \dots, S_n) \\ &= \max_{S_1, \dots, S_{n-1}} \prod_{t=1}^n P(y_t | S_t) P(S_t | S_{t-1}) \\ &= P(y_n | S_n) \max_{S_{n-1}} P(S_n | S_{n-1}) \gamma_{n-1}(S_{n-1}) \\ \log \gamma_n(S_n) &= \max_{S_1, \dots, S_{n-1}} \sum_{t=1}^n \left[ \log P(y_t | S_t) + \log P(S_t | S_{t-1}) \right] \\ &= \log P(y_n | S_n) + \max_{S_{n-1}} \left[ \log P(S_n | S_{n-1}) + \log \gamma_{n-1}(S_{n-1}) \right]\end{aligned}$$

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## Computing the Likelihood: The Forward Algorithm

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$$P(y_1, \dots, y_N) = \sum_{S_1, \dots, S_N} P(y_1, \dots, y_N, S_1, \dots, S_N)$$

## Computing the Likelihood: The Forward Algorithm

---

$$P(y_1, \dots, y_N) = \sum_{S_1, \dots, S_N} P(y_1, \dots, y_N, S_1, \dots, S_N)$$

$$P(y_1, \dots, y_N) = \sum_{S_N} \alpha_N(S_N)$$

$$\alpha_n(S_n) = P(y_1, \dots, y_n, S_n)$$

$$= \sum_{S_1} \dots \sum_{S_{n-1}} P(y_1, \dots, y_n, S_1, \dots, S_{n-1}, S_n)$$

$$= \sum_{S_1} \dots \sum_{S_{n-1}} \prod_{t=1}^n P(y_t | S_t) P(S_t | S_{t-1})$$

$$= \sum_{S_1} \dots \sum_{S_{n-1}} P(y_n | S_n) P(S_n | S_{n-1}) \prod_{t=1}^{n-1} P(y_t | S_t) P(S_t | S_{t-1})$$

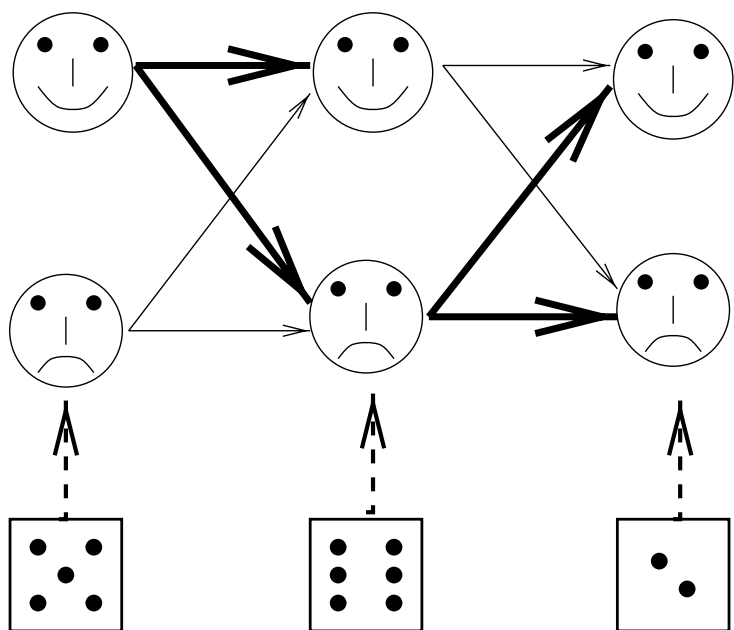
$$= P(y_n | S_n) \sum_{S_{n-1}} P(S_n | S_{n-1}) \sum_{S_1} \dots \sum_{S_{n-2}} \prod_{t=1}^{n-1} P(y_t | S_t) P(S_t | S_{t-1})$$

$$= P(y_n | S_n) \sum_{S_{n-1}} P(S_n | S_{n-1}) \alpha_{n-1}(S_{n-1})$$

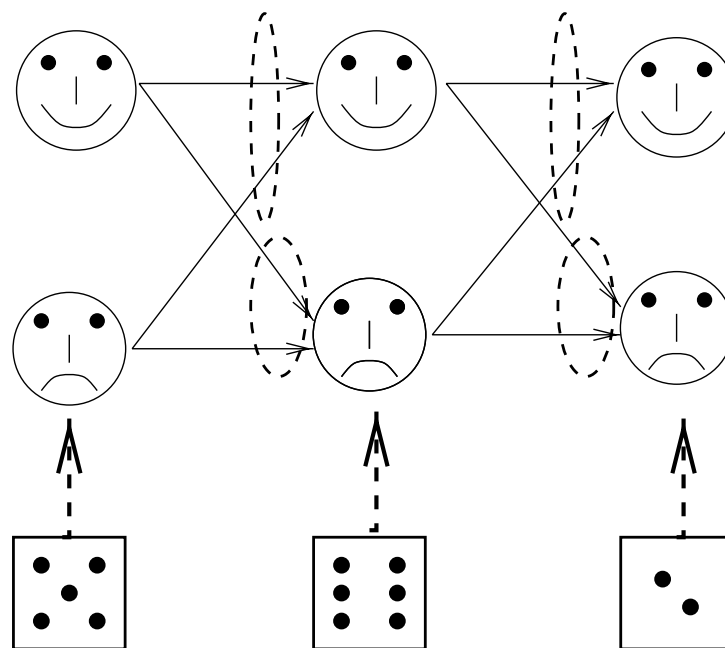
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## Comparison between the Viterbi and the Forward Algorithms

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Viterbi Algorithm:  
Maximisation



Forward Algorithm:  
Summation

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