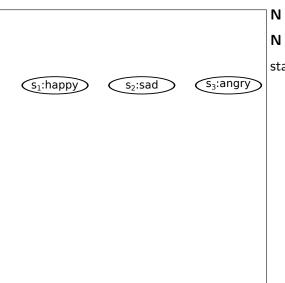
Let's consider a simple example:
a robot that tracks the emotional states of a player.

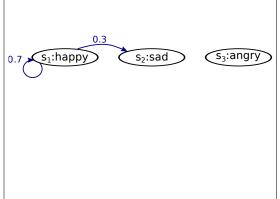


N - number of states

N = 3

states:

- *s*<sub>1</sub>: happy
- *s*<sub>2</sub>: sad
- *s*<sub>3</sub>: angry



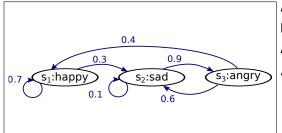
**A** - state transition probability distribution

$$\mathbf{A} = \{a_{i,j}\}, \ 1 \le i, j \le N$$
$$a_{i,j} = P(q_{t+1} = s_j | q_t = s_i)$$

- $a_{1,1} = 0.7$
- $a_{1,2} = 0.3$
- $a_{1,3} = 0$

$$\sum_{j=1}^{N} a_{i,j} = 1, \quad 1 \le i \le N$$

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**A** - state transition probability distribution

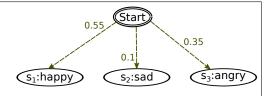
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$$\mathbf{A} = \begin{array}{cccc} s_1 & s_2 & s_3 \\ s_1 & 0.7 & 0.3 & 0 \\ 0 & 0.9 & 0.1 \\ s_3 & 0.4 & 0.6 & 0 \end{array}$$





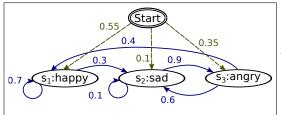
 $\Pi$  - initial state distribution

$$\Pi = {\pi_i}, \quad 1 \le i \le N$$

$$\pi_i = P(q_1 = s_i)$$

$$s_1 \quad s_2 \quad s_3$$

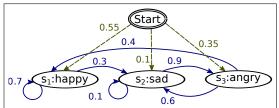
$$\Pi = (0.35 \quad 0.1 \quad 0.55)$$



$$A = \begin{array}{cccc} s_1 & s_2 & s_3 \\ s_1 & 0.7 & 0.3 & 0 \\ 0 & 0.9 & 0.1 \\ s_3 & 0.4 & 0.6 & 0 \end{array}$$

$$\Gamma = \begin{pmatrix} s_1 & s_2 & s_3 \\ 0.35 & 0.1 & 0.55 \end{pmatrix}$$

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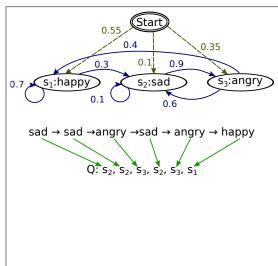


$$sad \rightarrow sad \rightarrow angry \rightarrow sad \rightarrow angry \rightarrow happy$$

$$A = \begin{array}{cccc} s_1 & s_2 & s_3 \\ s_1 & 0.7 & 0.3 & 0 \\ 0 & 0.9 & 0.1 \\ s_3 & 0.4 & 0.6 & 0 \end{array}$$

$$\Gamma = \begin{pmatrix} s_1 & s_2 & s_3 \\ 0.35 & 0.1 & 0.55 \end{pmatrix}$$

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$$A = \begin{cases} s_1 & s_2 & s_3 \\ 0.7 & 0.3 & 0 \\ 0 & 0.9 & 0.1 \\ 0.4 & 0.6 & 0 \end{cases}$$

$$A = \begin{cases} s_1 & s_2 & s_3 \\ 0.35 & 0.1 & 0.55 \end{cases}$$

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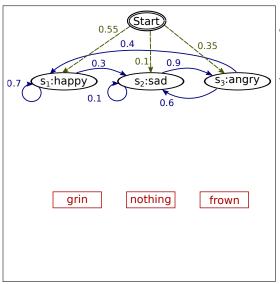
$$A = \begin{cases} 0.75 & 0.1 \\ 0.75 & 0.1 \end{cases}$$

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$$A = \begin{cases} 0.75 & 0.1 \\ 0.75 & 0.1 \end{cases}$$

$$A = \begin{cases} 0.7$$

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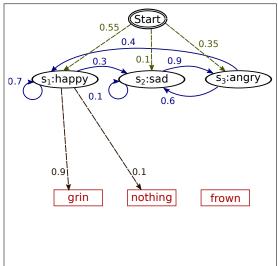


**M** - number of distinct observable values

$$M = 3$$

values:

- *v*<sub>1</sub>: grin
- $v_2$ : nothing
- *v*<sub>3</sub>: frown



**B** - observation values probability distribution

$$\mathbf{B} = \{b_{j,k}\} \ 1 \leq j \leq N, 1 \leq k, \leq M$$

$$b_{j,k} = b_j(v_k)$$
  
=  $P(o_t = v_k | q_t = s_j)$ 

• 
$$b_{1,1} = b_1(grin) = 0.9$$

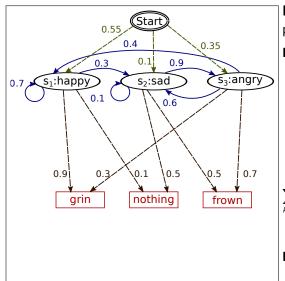
• 
$$b_{1,2} = b_1(nothing) = 0.1$$

• 
$$b_{1,3} = b_1(frown) = 0$$

$$\sum_{k=1}^{M} b_{j,k} = 1, \quad 1 \le j \le N$$



October 26, 2012



**B** - observation values probability distribution

$$\mathbf{B} = \{b_{j,k}\} \text{ } 1 \leq j \leq N, 1 \leq k, \leq M$$

$$b_{j,k} = b_j(v_k)$$
  
=  $P(o_t = v_k | q_t = s_j)$ 

• 
$$b_{1,1} = b_1(grin) = 0.9$$

• 
$$b_{1,2} = b_1(nothing) = 0.1$$

• 
$$b_{1,3} = b_1(frown) = 0$$

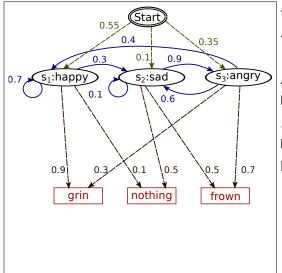
grin

$$\sum_{k=1}^{M} b_{j,k} = 1, \quad 1 \le j \le N$$

$$\mathbf{B} = \begin{array}{c} s_1 \\ s_2 \\ s_3 \end{array} \begin{pmatrix} 0.9 & 0.1 & 0 \\ 0 & 0.5 & 0.5 \\ 0.3 & 0 & 0.7 \end{array}$$

notg

frown



 $\lambda$  - parameters of the model

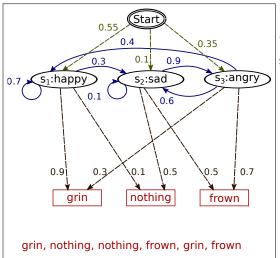
$$\lambda = (A, B, \Pi)$$

A - state transition probability distribution

*B* - observation values probability distribution

 $\Pi$  - initial state distribution

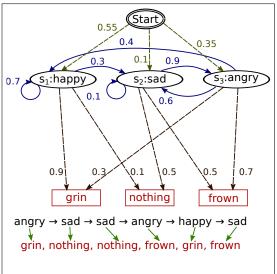
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O - observation sequence

**T** - length of observation sequence

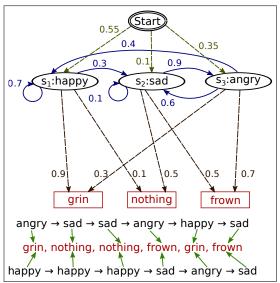
$$O = [o_1 o_2 \cdots o_T]$$



O - observation sequence

**T** - length of observation sequence

$$O = [o_1 o_2 \cdots o_T]$$



 ${f O}$  - observation sequence

**T** - length of observation sequence

$$O = [o_1 o_2 \cdots o_T]$$

• Example inspired from:

R. Zubek (2006). "Introduction to hidden markov models". In: *AI Game Programming Wisdom* 3, pp. 633–646

# Restating the three fundamental HMM Problems

#### Estimation

... luam de la Alex

$$P(O|Q,\lambda) = \prod_{t=1}^{T} P(o_t|q_t,\lambda) = \prod_{t=1}^{T} b_{q_t}(o_t) = b_{q_1}(o_1) \cdot \ldots \cdot b_{q_T}(o_T) \quad (1)$$

$$P(Q|\lambda) = \pi_{q_1} \prod_{t=2}^{I} a_{q_{t-1},q_t} = \pi_{q_1} \cdot a_{q_1,q_2} \cdot a_{q_2,q_3} \cdot \ldots \cdot a_{q_{T-1},q_T}$$
 (2)

$$P(O|\lambda) = \sum_{\text{all } Q} P(O, Q|\lambda) = \sum_{\text{all } Q} P(O, |Q, \lambda) \cdot P(Q, \lambda)$$

$$= \sum_{\text{all } Q} \left( \pi_{q_1} \cdot b_{q_1}(o_1) \cdot \prod_{t=2}^{T} b_{q_t}(o_t) a_{q_{t-1}, q_t} \right)$$
(3)