

# Natural Language Processing

## Lecture 11: Hidden Markov Models

# Finding POS Tags

Bill directed    plays    about    English kings

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

		p(t  Bill)
PropN	41	0.118
Verb	2	0.006
Noun	30 3	0.870

		p(t  directed )
Adj	0	0.000
Verb	1 0	1.000

		p(t  plays )
Verb	1 8	0.750
PIN	6	0.250

		p(t  about )
Prep	154 6	0.750
Adv	502	0.244
Part	12	0.006

# Running Example: POS

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

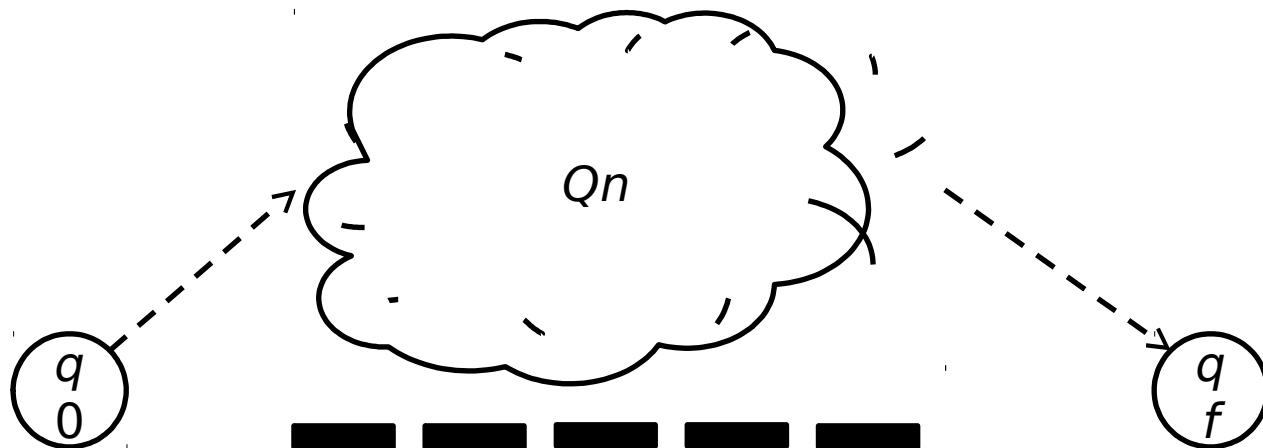
Adj  
Noun

PIN  
Verb

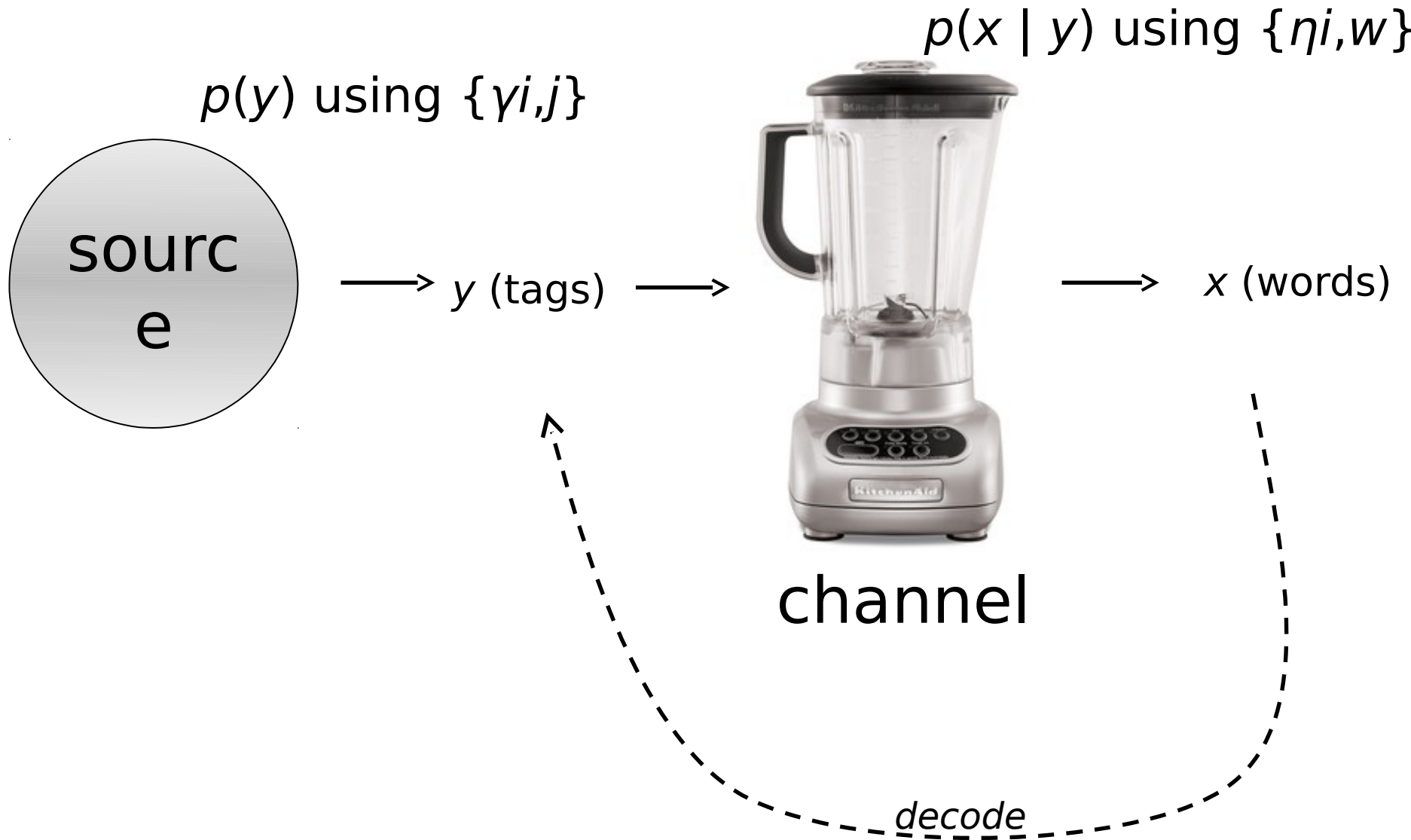
		$p(t \mid \text{English})$			$p(t \mid \text{kings})$
Adj	11	0.344	PIN	3	1.000
Noun	21	0.656	Verb	0	0.000

# Hidden Markov Model

- $q_0$ : start state (“silent”)
- $q_f$ : final state (“silent”)
- $Q$ : set of “normal” states (excludes  $q_0$  and final  $q_f$ )
- $\Sigma$ : vocabulary of observable symbols
- $\gamma_{i,j}$ : probability of transitioning to  $q_j$  given current state  $q_i$
- $\eta_{i,w}$ : probability of emitting  $w \in \Sigma$  given current state  $q_i$



# HMM as a Noisy Channel



# **States vs. Tags**



# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

$p(\text{PropN} \mid \langle S \rangle \langle S \rangle)$	0.202
$p(\text{Verb} \mid \langle S \rangle \langle S \rangle)$	0.023
$p(\text{Noun} \mid \langle S \rangle \langle S \rangle)$	0.040

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

p(PropN   <S> <S>)	0.202	p(Adj   <S> PropN)	0.004	0.00081
		p(Verb   <S> PropN)	0.139	0.02808
p(Verb   <S> <S>)	0.023	p(Adj   <S> Verb)	0.062	0.00143
		p(Verb   <S> Verb)	0.032	0.00074
p(Noun   <S> <S>)	0.040	p(Adj   <S> Noun)	0.005	0.00020
		p(Verb   <S> Noun)	0.222	0.00888

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

p(Adj   <S> PropN)	0.00081	p(Verb   PropN Adj)	0.011	0.00001
		p(PIN   PropN Adj)	0.157	0.00013
p(Verb   <S> PropN)	0.02808	p(Verb   PropN Verb)	0.162	<b>0.00455</b>
		p(PIN   PropN Verb)	0.022	0.00062
p(Adj   <S> Verb)	0.00143	p(Verb   Verb Adj)	0.009	0.00001
		p(PIN   Verb Adj)	0.246	0.00035
p(Verb   <S> Verb)	0.00074	p(Verb   Verb Verb)	0.078	0.00006
		p(PIN   Verb Verb)	0.034	0.00003
p(Adj   <S> Noun)	0.00020	p(Verb   Noun Adj)	0.020	0.00000
		p(PIN   Noun Adj)	0.103	0.00002
p(Verb   <S> Noun)	0.00888	p(Verb   Noun Verb)	0.176	0.00156
		p(PIN   Noun Verb)	0.018	0.00016

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

		$p(t \mid \text{Bill})$	$p(\text{Bill} \mid t)$
PropN	41	0.118	0.00044
Verb	2	0.006	0.00002
Noun	303	0.870	0.00228

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

		$p(t \mid \text{directed})$	$p(\text{directed} \mid t)$
Adj	0	0.000	0.00000
Verb	10	1.000	0.00008

# Running Example

Bill directed plays about English kings

PropN  
Verb  
Noun

Adj  
Verb

Verb  
PIN

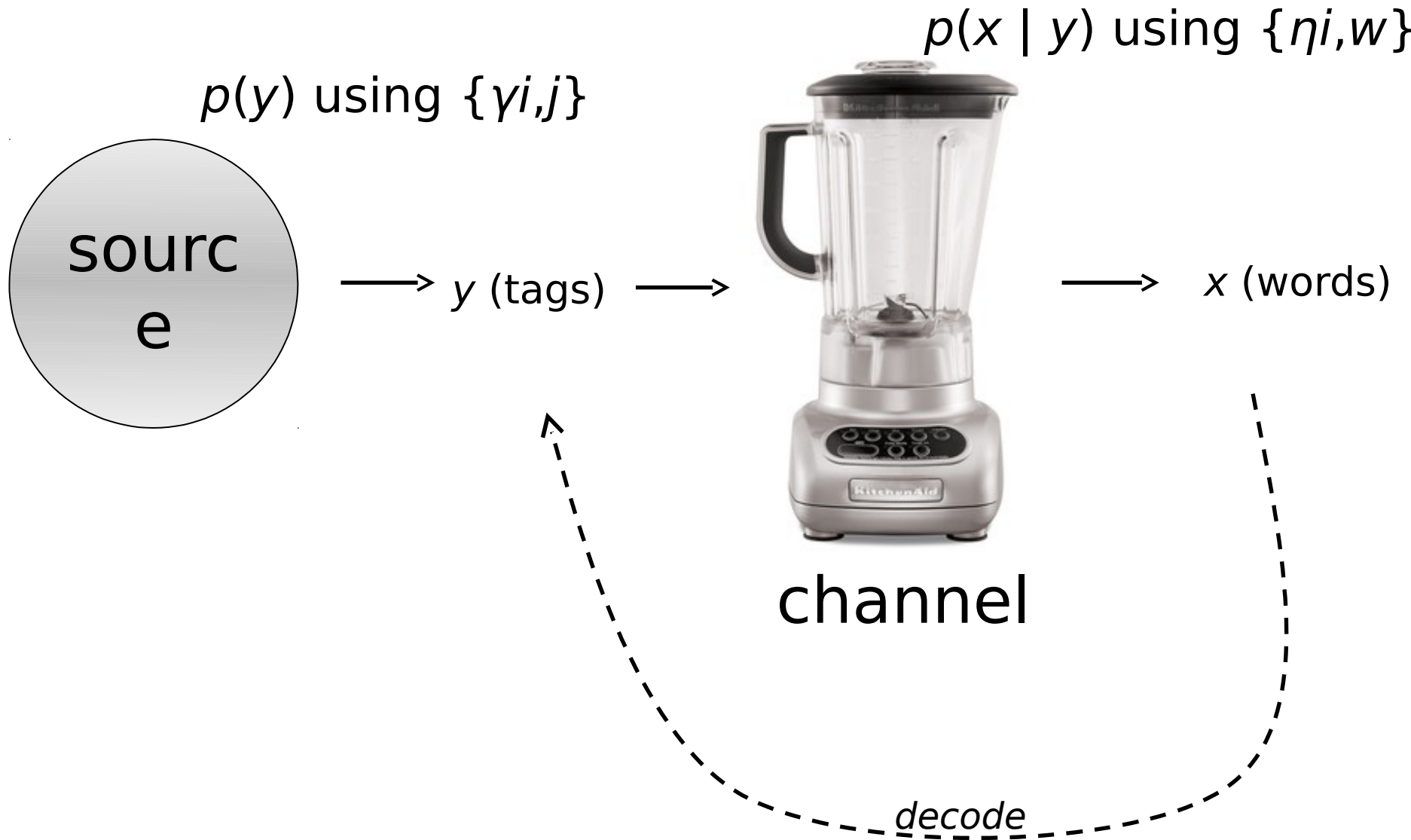
Prep  
Adv  
Part

Adj  
Noun

PIN  
Verb

		$p(t \mid \text{plays})$	$p(\text{plays} \mid t)$
Verb	18	0.750	0.00014
PIN	6	0.250	0.00010

# HMM as a Noisy Channel



# Part-of-Speech Tagging Task

- Input: a sequence of word tokens  $\mathbf{x}$
- Output: a sequence of part-of-speech tags  $\mathbf{y}$ , one per word

HMM solution: find the most likely tag sequence, given the word sequence.



If I knew the best state sequence for words  $x_1 \dots x_{n-1}$ , then I could figure out the last state.

That decision would depend only on state  $n-1$ .

$$\begin{aligned} y_n^* &= \arg \max_{q_i \in Q} p(Y_1 = y_1^*, \dots, Y_{n-1} = y_{n-1}^*, Y_n = q_i \mid \mathbf{x}) \\ &= \arg \max_{q_i \in Q} V[n-1, y_{n-1}^*] \cdot \gamma_{y_{n-1}^*, i} \cdot \eta_{i, x_n} \cdot \gamma_{i, f} \\ &= \arg \max_{q_i \in Q} \gamma_{y_{n-1}^*, i} \cdot \eta_{i, x_n} \cdot \gamma_{i, f} \end{aligned}$$

I don't know that best sequence, but there are only  $|Q|$  options at  $n-1$ .

So I only need the score of the best sequence up to  $n-1$ , ending in each possible state at  $n-1$ . Call this  $V[n-1, q]$  for  $q \in Q$ .

Ditto, at every other timestep  $n-2, n-3, \dots 1$ .

# Viterbi Algorithm (Recursive Equations)

$$V[0, q_0] = 1$$

$$V[t, q_j] = \max_{q_i \in Q \cup \{q_0\}} V[t - 1, q_i] \cdot \gamma_{i,j} \cdot \eta_{j,x_t}$$

$$\text{goal} = \max_{q_i \in Q} V[n, q_i] \cdot \gamma_{i,f}$$

# Viterbi Algorithm (Procedure)

$V[*, *] \leftarrow 0$

$V[0, q_0] \leftarrow 1$

for  $t = 1 \dots n$

  foreach  $q_j$

    foreach  $q_i$

$V[t, q_j] \leftarrow \max\{V[t, q_j], V[t - 1, q_i] \times \gamma_{i,j} \times \eta_{i,x_t}\}$

  foreach  $q_i$

    goal  $\leftarrow \max\{ \text{goal}, V[n, q_i] \times \gamma_{i,f} \}$

return goal

# Running Example

Bill directed plays about English kings

$q_0$	1							
$q_1$								
$q_2$								
$q_3$								
$q_4$								
...								
$q_i$								
$Q_i$								
$q_f$								

# Running Example

Bill directed plays about English kings

