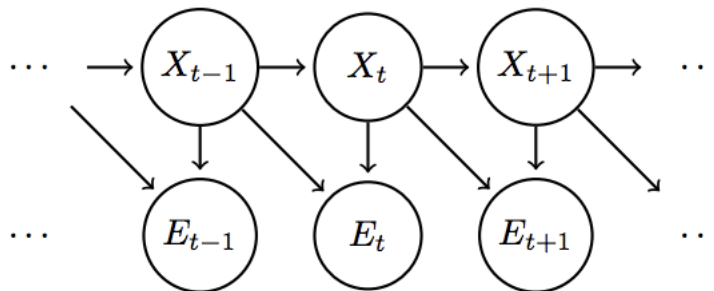


## 1 HMMs: Warmup

1. What are the three components of a hidden markov model? What makes it "hidden"?
  
2. Write an expression for the joint distribution of a hidden markov model consisting of states  $X_0, \dots, X_n$  and evidence variables  $E_1, \dots, E_N$ . How does the expression reflect the underlying structure of the model?
  
3. For each of the following descriptions in English of an inference task, write the corresponding probability expression:
  - Draw conclusions about our current underlying state given evidence up to the current time step
  - Draw conclusions about our future underlying state given evidence up to the current time step
  - Draw conclusions about a past underlying state given evidence up to the current time step
  - Draw conclusions about the sequence of underlying states given evidence up to the current time step
  - Draw conclusions about the most likely sequence of underlying states given evidence up to the current time step

4. Hidden Markov Models can be extended in a number of ways to incorporate additional relations. Since the independence assumptions are different in these extended Hidden Markov Models, the forward algorithm updates will also be different. What is the forward algorithm updates for the extended Hidden Markov Models specified by the following Bayes net?



## 2 HMMs: Tracking a Jabberwock

You have been put in charge of a Jabberwock for your friend Lewis. The Jabberwock is kept in a large tugley wood which is conveniently divided into an  $N \times N$  grid. It wanders freely around the  $N^2$  possible cells. At each time step  $t = 1, 2, 3, \dots$ , the Jabberwock is in some cell  $X_t \in \{1, \dots, N\}^2$ , and it moves to cell  $X_{t+1}$  randomly as follows: with probability  $1 - \epsilon$ , it chooses one of the (up to 4) valid neighboring cells uniformly at random; with probability  $\epsilon$ , it uses its magical powers to teleport to a random cell uniformly at random among the  $N^2$  possibilities (it might teleport to the same cell). Suppose  $\epsilon = \frac{1}{2}$ ,  $N = 10$  and that the Jabberwock always starts in  $X_1 = (1, 1)$ .

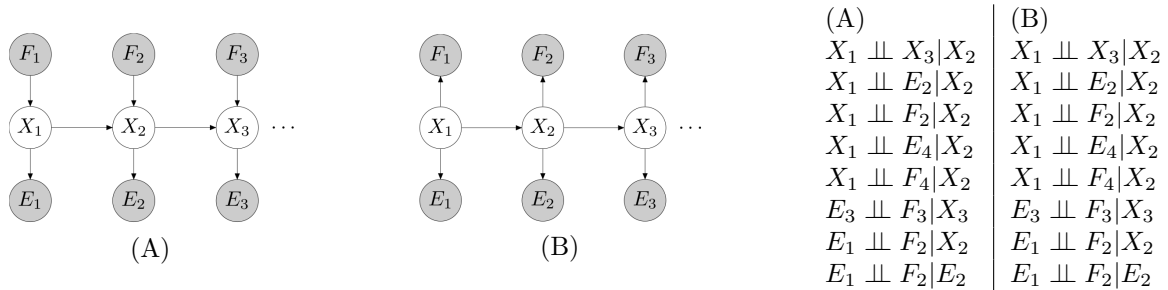
- (a) Compute the probability that the Jabberwock will be in  $X_2 = (2, 1)$  at time step 2. What about  $P(X_2 = (4, 4))$ ?

At each time step  $t$ , you don't see  $X_t$  but see  $E_t$ , which is the row that the Jabberwock is in; that is, if  $X_t = (r, c)$ , then  $E_t = r$ . You still know that  $X_1 = (1, 1)$ .

- (b) Suppose we see that  $E_1 = 1$ ,  $E_2 = 2$ . Fill in the following table with the distribution over  $X_t$  after each time step, taking into consideration the evidence. Your answer should be concise. Hint: you should not need to do any heavy calculations.

$t$	$P(X_t   e_{1:t-1}, X_1 = (1, 1))$		$P(X_t   e_{1:t}, X_1 = (1, 1))$	
1	$X_1$	$P(X_1)$	$X_1$	$P(X_1)$
	(1, 1)		(1, 1)	
	all other values		all other values	
2	$X_2$	$P(X_2   e_1, X_1 = (1, 1))$	$X_2$	$P(X_2   e_{1:2}, X_1 = (1, 1))$
	(1, 2)		(2, 1)	
	(2, 1)		(2, $a$ ) ( $\forall a, a > 1$ )	
	all other values		all other values	

You are a bit unsatisfied that you can't pinpoint the Jabberwock exactly. But then you remembered Lewis told you that the Jabberwock teleports only because it is frumious on that time step, and it becomes frumious independently of anything else. Let us introduce a variable  $F_t \in \{0, 1\}$  to denote whether it will teleport at time  $t$ . We want to add these frumious variables to the HMM. Consider the two candidates:



(c) For each model, circle the conditional independence assumptions above which are true in that model.

(d) Which Bayes net is more appropriate for the problem domain here, (A) or (B)? Justify your answer.

For the following questions, your answers should be fully general for models of the structure shown above, not specific to the teleporting Jabberwock.

(e) For (A), express  $P(X_{t+1}, e_{1:t+1}, f_{1:t+1})$  in terms of  $P(X_t, e_{1:t}, f_{1:t})$  and the conditional probability tables used to define the network. Assume the  $E$  and  $F$  nodes are all observed.

(f) For (B), express  $P(X_{t+1}, e_{1:t+1}, f_{1:t+1})$  in terms of  $P(X_t, e_{1:t}, f_{1:t})$  and the CPTs used to define the network. Assume the  $E$  and  $F$  nodes are all observed.

Suppose that we don't actually observe the  $F_t$ s.

(g) For (A), express  $P(X_{t+1}, e_{1:t+1})$  in terms of  $P(X_t, e_{1:t})$  and the CPTs used to define the network.

(h) For (B), express  $P(X_{t+1}, e_{1:t+1})$  in terms of  $P(X_t, e_{1:t})$  and the CPTs used to define the network.