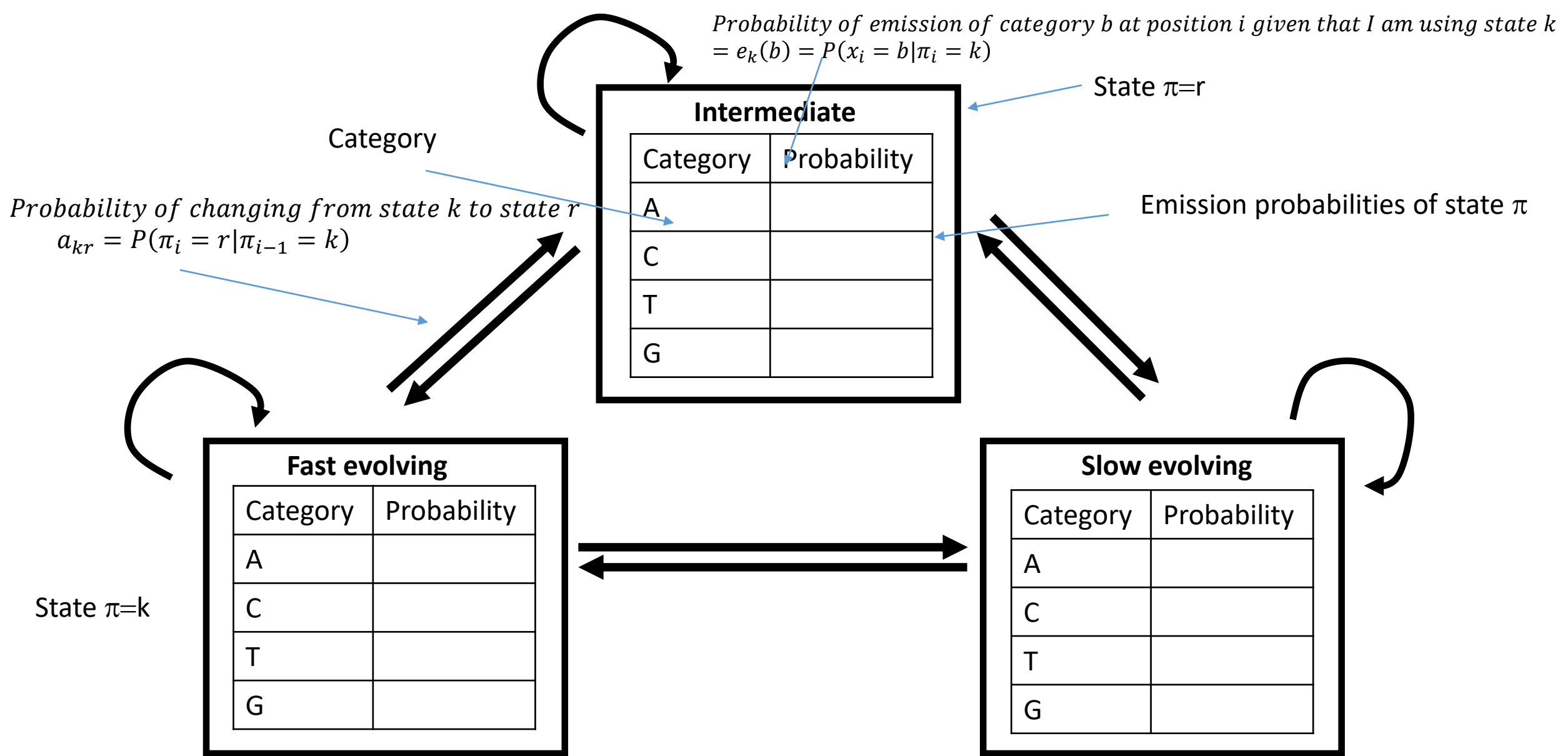
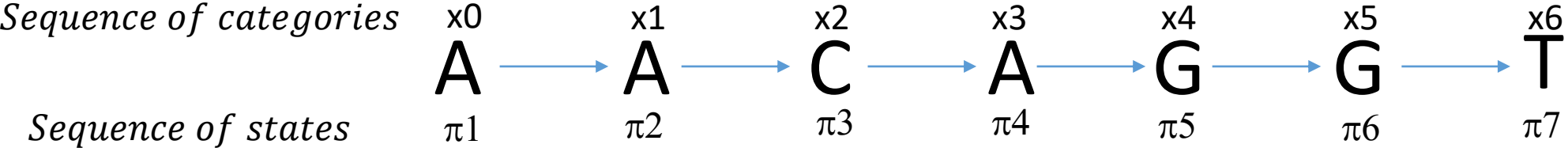


# Practical Session 2

MM, HMM and sequences

REMEMBER!



# NOTATION HMM

*State at position  $i = \pi_i$*

*Sequence of states  $= \pi$*

*Category at position  $i = x_i$*

*Sequence of categories  $= x$*

$$P(x, \pi) = a_{0\pi} \prod_i^N e_{\pi_i}(x_i) a_{\pi_i \pi_{i+1}}$$



Initial state ("Probability that I started  
the chain at state  $\pi$ ")

Probability of changing from state  $k$  to state  $r = a_{kr} = P(\pi_i = r | \pi_{i-1} = k)$

I am using state  $r$

Given that I was in the previous position using state  $k$

Transmission matrix

	Fast Evolving	Intermediate Evolving	Slow Evolving
Fast Evolving			
Intermediate Evolving			
Slow Evolving			

Emission Probability matrix for each possible state

Fast evolving		Intermediate		Slow evolving	
Category	Probability	Category	Probability	Category	Probability
A		A		A	
C		C		C	
T		T		T	
G		G		G	

Probability of emission of category  $b$  at position  $i$  given that I am using state  $k = e_k(b) = P(x_i = b | \pi_i = k)$

# HMM: Probability of $x$ given model? Forward

## $P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 | \text{HMM})$

	1	3	4	...
F	$0.5 * 1/6$			
L	$0.5 * 1/10$			

$x$	$P(x F)$		$x$	$P(x C)$
1	$\frac{1}{6}$	$P(C F) = 0.01$	1	$\frac{1}{10}$
2	$\frac{1}{6}$		2	$\frac{1}{10}$
3	$\frac{1}{6}$		3	$\frac{1}{10}$
4	$\frac{1}{6}$		4	$\frac{1}{10}$
5	$\frac{1}{6}$	$P(F C) = 0.1$	5	$\frac{1}{10}$
6	$\frac{1}{6}$		6	$\frac{1}{2}$

# HMM: Probability of $x$ given model? Forward

$$P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 \mid \text{HMM})$$

	1	3	4	...
F	0.5*1/6	0.5*1/6*0.99*1/6		
L	0.5*1/10			

$x$	$P(x F)$
1	$\frac{1}{6}$
2	$\frac{1}{6}$
3	$\frac{1}{6}$
4	$\frac{1}{6}$
5	$\frac{1}{6}$
6	$\frac{1}{6}$

$P(C|F) = 0.01$   
 $\longrightarrow$

$x$	$P(x C)$
1	$\frac{1}{10}$
2	$\frac{1}{10}$
3	$\frac{1}{10}$
4	$\frac{1}{10}$
5	$\frac{1}{10}$
6	$\frac{2}{10}$

$\longleftarrow$   
 $P(F|C) = 0.1$

# HMM: Probability of $x$ given model? Forward

## $P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 | \text{HMM})$

	1	3	4	...
F	$0.5 \cdot 1/6$	$0.5 \cdot 1/6 \cdot 0.99 \cdot 1/6 + 0.5 \cdot 1/10 \cdot 0.1 \cdot 1/6$		
L	$0.5 \cdot 1/10$			

Diagram illustrating the forward algorithm for HMM. The table shows the probability of the sequence  $x$  given the model, calculated using the forward algorithm. The sequence  $x$  is  $1, 3, 4, 5, 1, 2, 6, 6, 3, 6, 6, 2, 1, 2, 4, 5$ . The model parameters are  $P(C|F) = 0.01$  and  $P(F|C) = 0.1$ . The table shows the probability of the sequence  $x$  given the model, calculated using the forward algorithm. The sequence  $x$  is  $1, 3, 4, 5, 1, 2, 6, 6, 3, 6, 6, 2, 1, 2, 4, 5$ . The model parameters are  $P(C|F) = 0.01$  and  $P(F|C) = 0.1$ .

$x$	$P(x F)$		$x$	$P(x C)$
1	$\frac{1}{6}$		1	$\frac{1}{10}$
2	$\frac{1}{6}$		2	$\frac{1}{10}$
3	$\frac{1}{6}$	$P(C F) = 0.01$	3	$\frac{1}{10}$
4	$\frac{1}{6}$		4	$\frac{1}{10}$
5	$\frac{1}{6}$	$P(F C) = 0.1$	5	$\frac{1}{10}$
6	$\frac{1}{6}$		6	$\frac{1}{2}$



# HMM: Probability of $x$ given model? Forward

$$P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 | \text{HMM})$$

F,F

C,F

	1	3	4	...
F	$0.5 \cdot 1/6$	$0.5 \cdot 1/6 \cdot 0.99 \cdot 1/6 + 0.5 \cdot 1/10 \cdot 0.1 \cdot 1/6$		
L	$0.5 \cdot 1/10$	$0.5 \cdot 1/10 \cdot 0.9 \cdot 1/10$		

C,C

$x$	$P(x F)$		$x$	$P(x C)$
1	$\frac{1}{6}$	$P(C F) = 0.01$	1	$\frac{1}{10}$
2	$\frac{1}{6}$		2	$\frac{1}{10}$
3	$\frac{1}{6}$		3	$\frac{1}{10}$
4	$\frac{1}{6}$		4	$\frac{1}{10}$
5	$\frac{1}{6}$	$P(F C) = 0.1$	5	$\frac{1}{10}$
6	$\frac{1}{6}$		6	$\frac{1}{2}$

# HMM: Probability of $x$ given model? Forward

$$P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 | \text{HMM})$$

	1	3		4	...
F	$0.5 \cdot 1/6$	$0.5 \cdot 1/6 \cdot 0.99 \cdot 1/6 + 0.5 \cdot 1/10 \cdot 0.1 \cdot 1/6$			
L	$0.5 \cdot 1/10$	$0.5 \cdot 1/10 \cdot 0.9 \cdot 1/10 + 0.5 \cdot 1/6 \cdot 0.01 \cdot 1/6$			

F,F  
C,F

C,C  
F,C

$x$	$P(x F)$		$x$	$P(x C)$
1	$\frac{1}{6}$		1	$\frac{1}{10}$
2	$\frac{1}{6}$		2	$\frac{1}{10}$
3	$\frac{1}{6}$	$P(C F) = 0.01$	3	$\frac{1}{10}$
4	$\frac{1}{6}$		4	$\frac{1}{10}$
5	$\frac{1}{6}$	$P(F C) = 0.1$	5	$\frac{1}{10}$
6	$\frac{1}{6}$		6	$\frac{1}{2}$

We have searched all the combinations

# HMM: Probability of x given model? Forward

## $P(1,3,4,5,1,2,6,6,3,6,6,2,1,2,4,5 | \text{HMM})$

	1	3	4	...
F	$0.5 \cdot 1/6$	$0.5 \cdot 1/6 \cdot 0.99 \cdot 1/6 + 0.5 \cdot 1/10 \cdot 0.1 \cdot 1/6$		
L	$0.5 \cdot 1/10$	$0.5 \cdot 1/10 \cdot 0.9 \cdot 1/10 + 0.5 \cdot 1/6 \cdot 0.01 \cdot 1/6$		

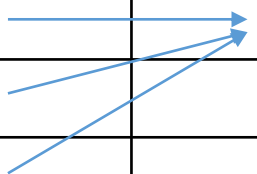
$P1$   
 $P2$

$P1 + P2 = P(S | \text{HMM})$

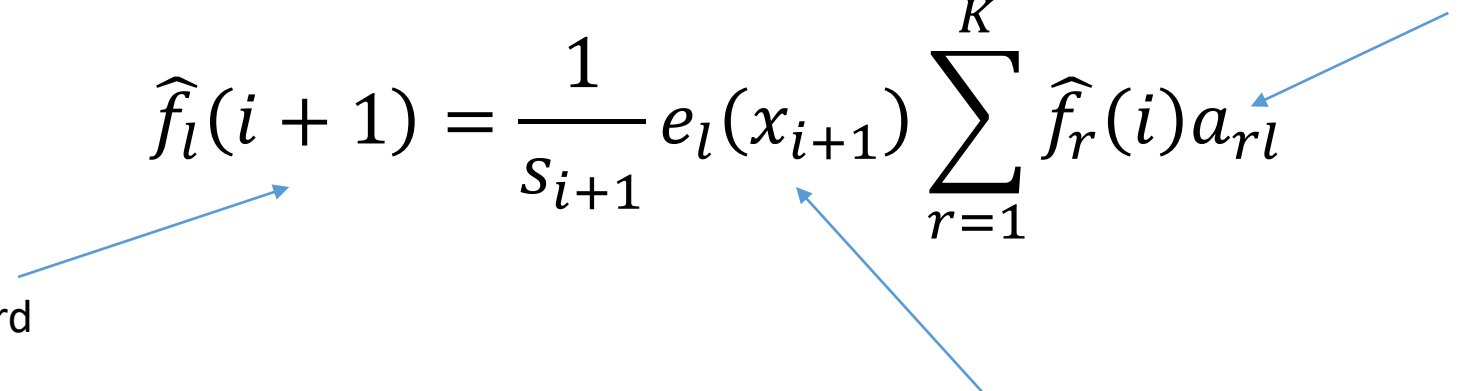
# HMM

- Scale the probability at each step (Rabiner 1989)

State	$\hat{f}_l(i)$	$\hat{f}_l(i + 1)$
State1		
State2		
StateK		



Probability of moving from state  $r$  to state  $l$  in  $i+1$

$$\hat{f}_l(i + 1) = \frac{1}{s_{i+1}} e_l(x_{i+1}) \sum_{r=1}^K \hat{f}_r(i) a_{rl}$$


Scaled forward

Emission probability of category at  $x_{i+1}$  using state  $l$

# HMM

- Scale the probability at each step (Rabiner 1989)

State	$\hat{f}_l(i)$	$\hat{f}_l(i + 1)$
State1		
State2		
StateK		

If we impose that

Then  $s_{i+1}$  is

Total of the scaled value

$$\sum_l \hat{f}_l(i) = 1$$

$$s_{i+1} = \sum_l e_l(x_{i+1}) \sum_k \hat{f}_k(i) a_{kl}$$

$$P(X|HMM) = \prod_i^L s_i$$

# Task 1

- Create in Python a class called HMM
- The constructor takes as input the transition probability matrix between states and a list of emission matrices, one for each state. Code these objects as a dictionary.

# Task 1

```
from AliasVose import RandomMultinomial
```

```
'''
```

```
A class to code methods that are used in HMM.
```

```
'''
```

```
class HiddenMarkovModel(object):
```

```
'''
```

```
Create an object HiddenMarkovModel with transition states and emission probabilities for each state
```

```
Transition is a dictionary where, for each state, we count the probability to move to another state
```

```
Emission is a dictionary where, for each state, we store the probabilities of each category
```

```
'''
```

```
def __init__(self, transition, emission):
```

```
    if transition.__len__() != emission.__len__():
```

```
        raise Exception("for each state, we must have an emission probability vector, but found " + transition. len () + " " + emission. len ())
```

```
    self.n = transition.__len__()
```

```
    self.transition = transition
```

```
    self.emission = emission
```

```
'''
```

```
dictionary to store the random multinomial
```

```
'''
```

```
self.random_transition = {}
```

```
self.random_emission = {}
```

## Task 2

- Implement a function that takes as input an observed sequence of categories and returns the log-likelihood. The scaled version is already implemented. Implement the unscaled version.



# Task 3

- Implement a method that returns a sequence and the hidden states given the transition and emission probabilities of the HMM