# Probabilistic Models for Biological Sequences

#### Laboratory of Bioinformatics I Module 2

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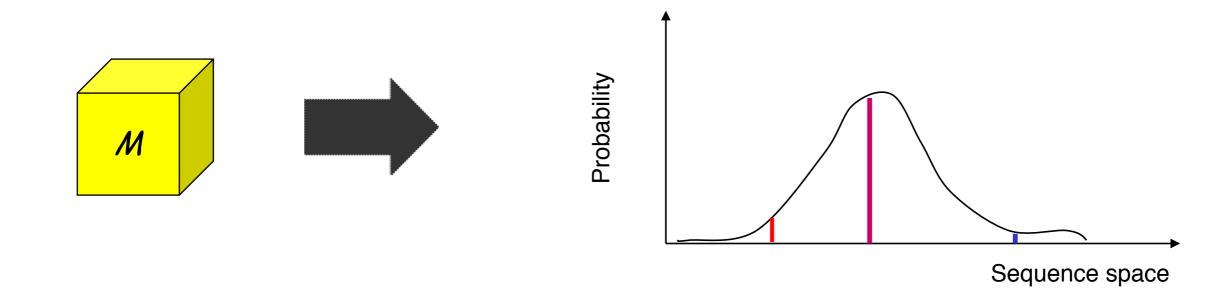
#### Models for Sequence

#### Generative definition:

- Objects producing different outcomes (sequences) with different probabilities
- The probability distribution over the sequences space determines the model specificity

Generates  $s_i$  with probability  $P(s_i \mid M)$ 

e.g.: *M* is the representation of the family of globins



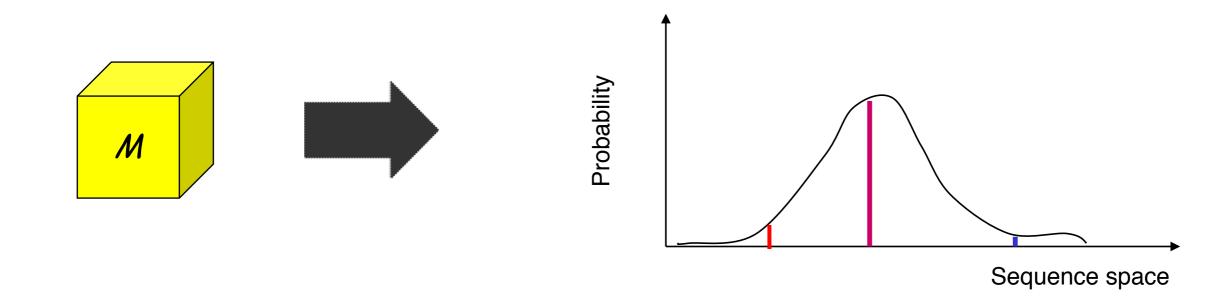
#### **Associative Definition**

The generative definition is useful as operative definition

Objects that, given an outcome (sequence), compute a probability value

Calculates the associated probability  $P(s_i \mid M)$  to  $s_i$ .

e.g.: *M* is the representation of the family of globins

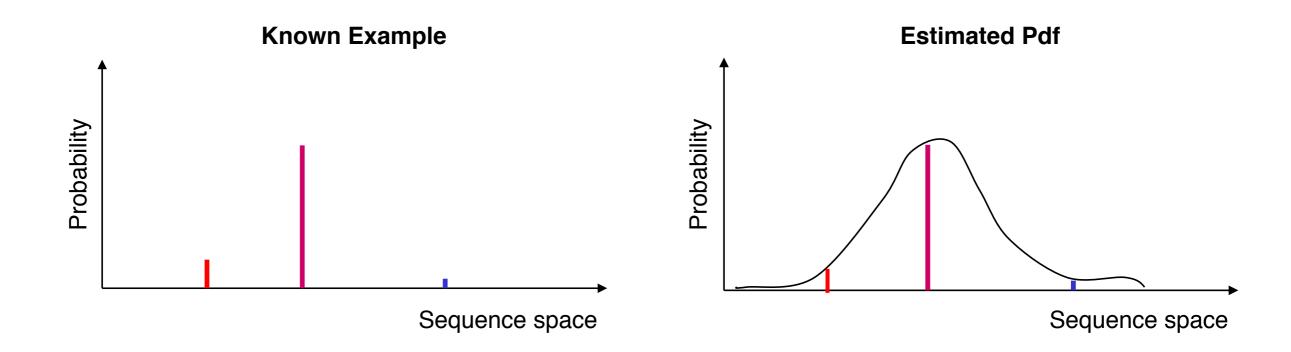


#### Which Model?

The most useful probabilistic models are Trainable systems

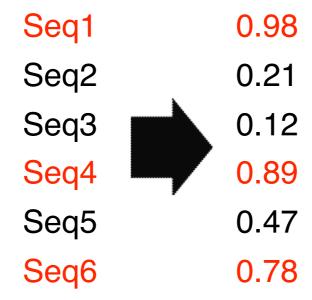
The probability density function over the sequence space can be estimated from known examples by a learning algorithm

Define a generic representation of the sequences of globins starting from a set of known globins



### Similarity Measure

Given a class of proteins (e.g. Globins), a probabilistic model trained on this family can be adopted to compute a probability value for new sequences



This value measures the similarity between the new sequence and the family described by the model

#### Which Probability?

A model M associates to a sequence  $s_i$  the probability  $P(s_i \mid M)$ This probability answers the question:

Which is the probability for a model M (e.g. describing the Globins) to generate the sequence  $s_i$ ?

The question we want to answer is:

Given a sequence  $s_i$ , does it belong to the class described by the model M? (e.g. is it a Globin?)

We need to compute  $P(M \mid s_i)$ 

#### **Bayes Theorem**

$$P(X,Y) = P(X \mid Y) P(Y) = P(Y \mid X) P(X)$$
 Joint probability

$$P(Y \mid X) = \frac{P(X \mid Y) P(Y)}{P(X)}$$

$$P(M \mid s_i) = \frac{P(s_i \mid M) P(M)}{P(s_i)}$$

P(M) and  $P(s_i)$ A priori probabilities

P(M) is the probability of the model (i.e. of the class described by the model) BEFORE we know the sequence:

Can be estimated as the abundance of the class

 $P(s_i)$  is the probability of the sequence in the sequence space.

Cannot be reliably estimated!!

# Comparing Models

We can overcome the problem comparing the probability of generating  $s_i$  from different models

$$\frac{P(M_1 \mid s_i)}{P(M_2 \mid s_i)} = \frac{P(s_i \mid M_1) P(M_1)}{P(s_i)} \frac{P(s_i)}{P(s_i \mid M_2) P(M_2)} = \frac{P(s_i \mid M_1) P(M_1)}{P(s_i \mid M_2) P(M_2)}$$

$$\frac{P(M_1)}{P(M_2)}$$

Ratio between the abundances of the classes

#### **Null Model**

Alternatively, we can score a sequence for a model M comparing it to a Null Model:

a model that generates ALL the possible sequences with probabilities depending ONLY on letter (e.g. residue) statistical abundance

$$S(M \mid s_i) = log \frac{P(s_i \mid M)}{P(s_i \mid N)}$$
Sequences NOT belonging to model M

Sequences belonging to model M

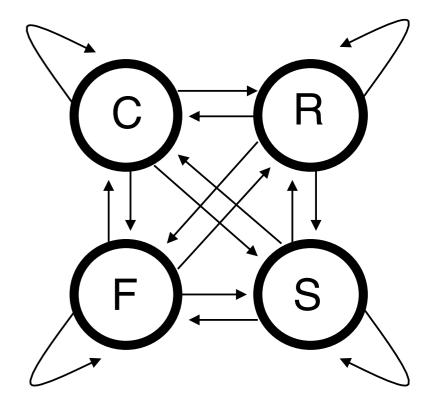
In this case we need a threshold and a statistic for evaluating the significance (E-value, P-value)

### A Simple Model

Time series of the weather conditions as a first hypothesis the weather condition in a day probabilistically depends ONLY on the weather conditions in the day before.

Define the conditional probabilities

P(CIC), P(CIR),.... P(RIC).....



The probability for the 5-days registration

**CRRCS** 

 $P(CRRCS) = P(C) \cdot P(RIC) \cdot P(RIR) \cdot P(CIR) \cdot P(SIC)$ 

C: Clouds

R: Rain

F: Fog

S: Sun

#### Markov Model

Stochastic generator of sequences in which the probability of state in position *i* depends ONLY on the state in position *i-1* 

Given a set of states (== alphabet)

$$C = \{c_1; c_2; c_3; \dots c_N \}$$

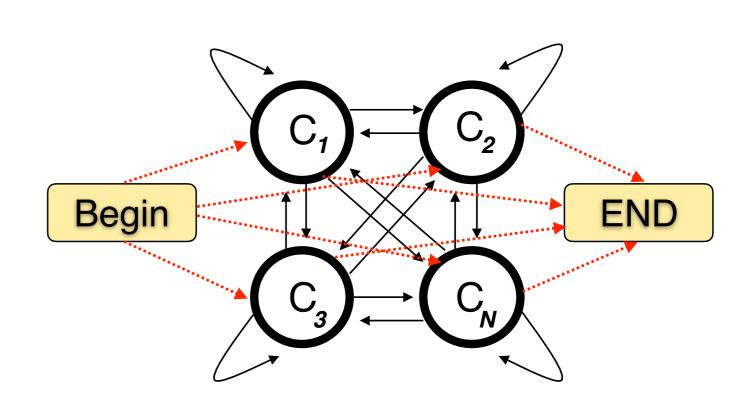
a Markov model is described with  $N \times (N+2)$  parameters

$$\{a_{r,t}, a_{BEGIN,t}, a_{r,END} \text{ with } r, t \in C\}$$

$$a_{r,q} = P(s_i = q \mid s_{i-1} = r)$$
 $a_{BEGIN,q} = P(s_1 = q)$ 
 $a_{r,END} = P(s_T = END \mid s_{T-1} = r)$ 

$$\sum_{t} a_{r,t} + a_{r,END} = 1 \ \forall \ r$$

$$\sum_{t} a_{BEGIN,t} = 1$$



# Sequence Probability

#### Given the sequence:

$$s = s_1 s_2 s_3 s_4 s_6 \dots s_T$$
 with  $s_i \in C = \{c_1; c_2; c_3; \dots c_N\}$ 

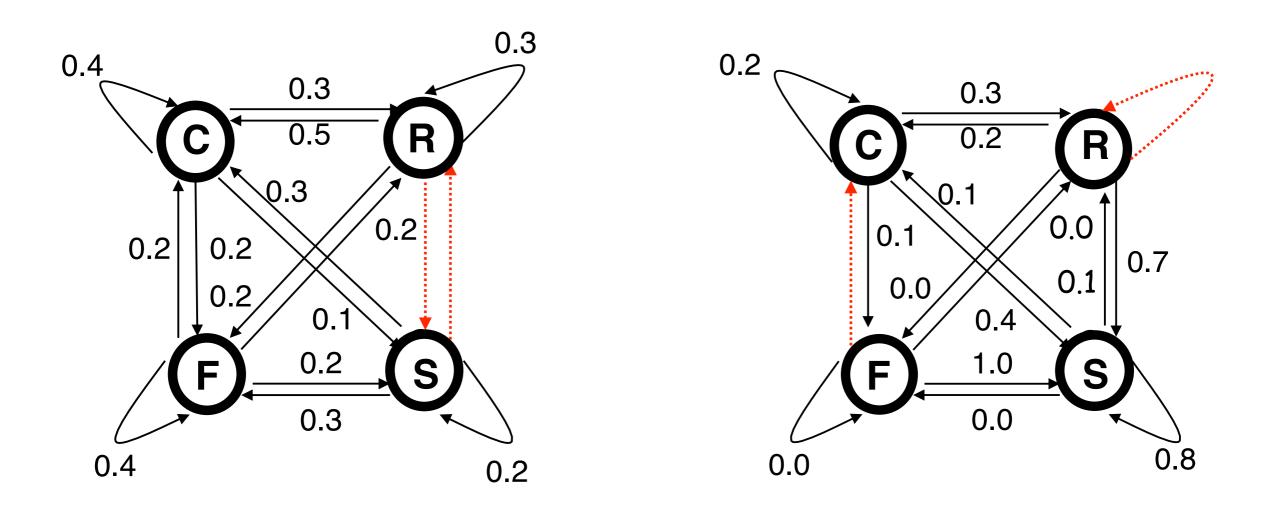
$$P(s \mid M) = P(s_1) \prod_{i=2}^{T} P(s_i \mid s_{i-1}) =$$

$$a_{BEGIN,s_1} \times \prod_{i=2}^{T} a_{s_{i-1},s_i} \times a_{s_{i-1},END}$$

$$P(ALKALI') = a_{BEGIN,A} \times a_{A,L} \times a_{L,K} \times a_{K,A} \times a_{A,L} \times a_{L,I} \times a_{L,END}$$

### **Probability Constrains**

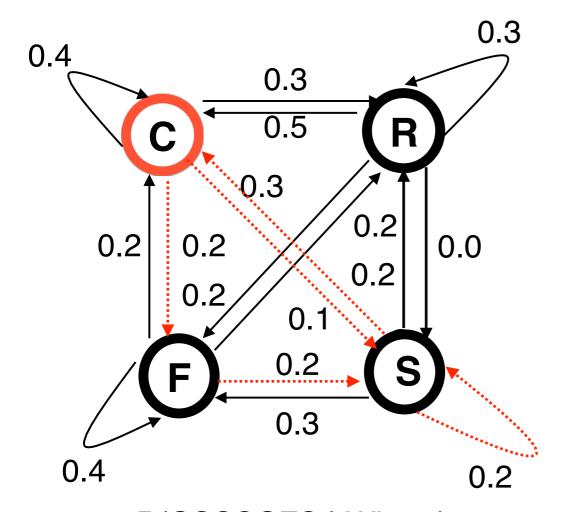
What are the missing probabilities given the constrains?



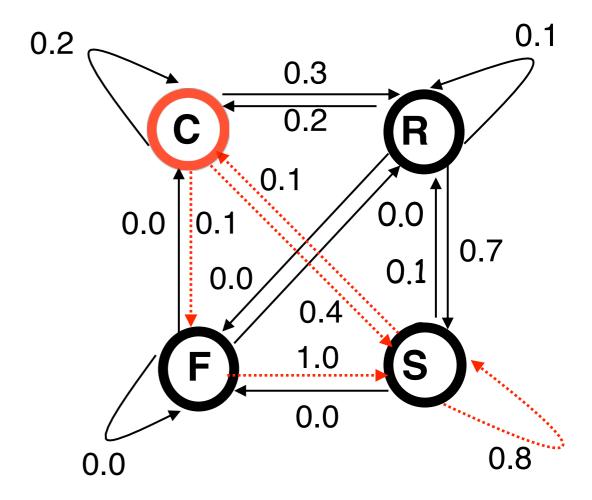
What is the better model to describe the weather in winter?

### **Probability Calculation**

Consider the sequence "CSSSCFS" and calculate its probability with both models when  $P(X \mid BEGIN) = 0.25$ 



P(CSSSCFS | Winter) = 0.25×0.1×0.2×0.2×0.3×0.2×0.2



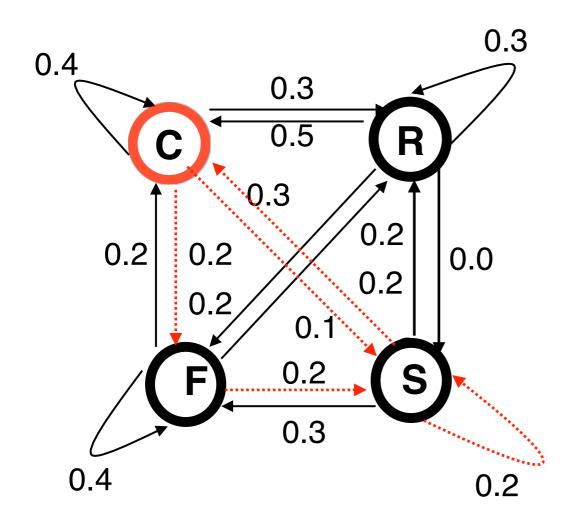
P(CSSSCFS | Summer) = 0.25×0.4×0.8×0.8×0.1×0.1×1.0

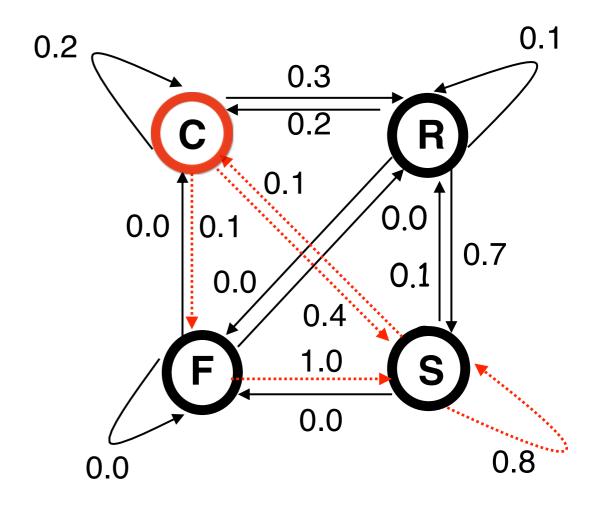
To which season the weather sequence is more likely to belong?

### **Probability Calculation**

P (Seq I Winter) =  $1.2 \times 10^{-5}$ 







$$\frac{P(Summer | Seq)}{P(Winter | Seq)} = \frac{P(Seq | Summer)}{P(Seq | Winter)} \times \frac{P(Summer)}{P(Winter)}$$

$$\times \frac{P(Summer)}{P(Winter)}$$
with 
$$\frac{P(Summer)}{P(Winter)} \approx 1$$