

Example: HMMs for CpG island finding

Ben Langmead

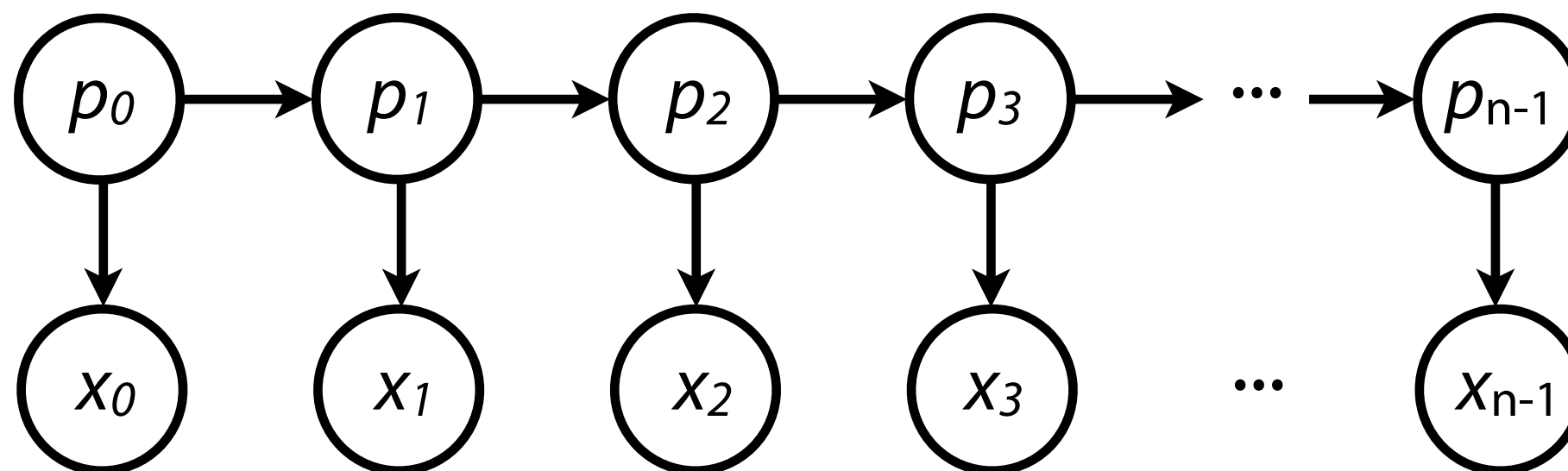


For original Keynote files, email me (ben.langmead@gmail.com)

Hidden Markov Model

We know what it is, how to calculate a joint probability, and how to find the most likely path given an emission sequence (Viterbi)

Can we design an HMM for finding **CpG islands**?

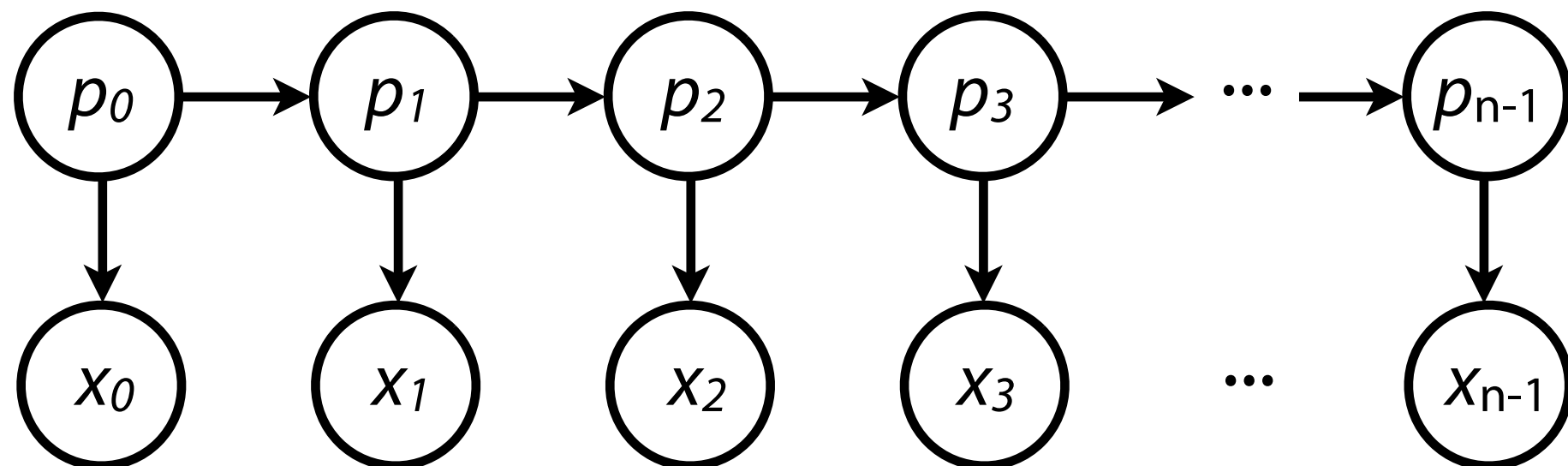


Hidden Markov Model

Idea 1:

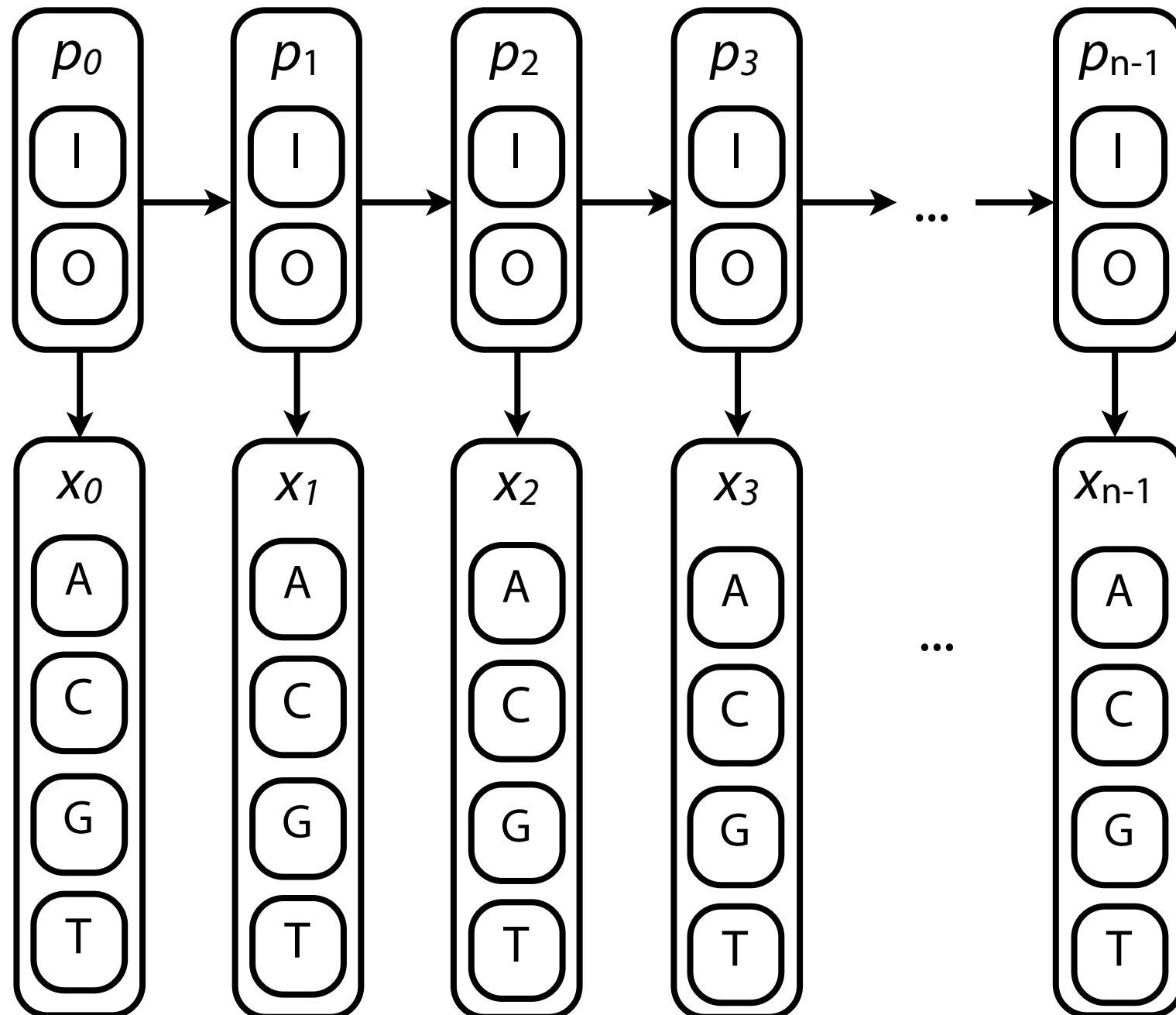
States (Q) = { }

Emissions (Σ) = { }



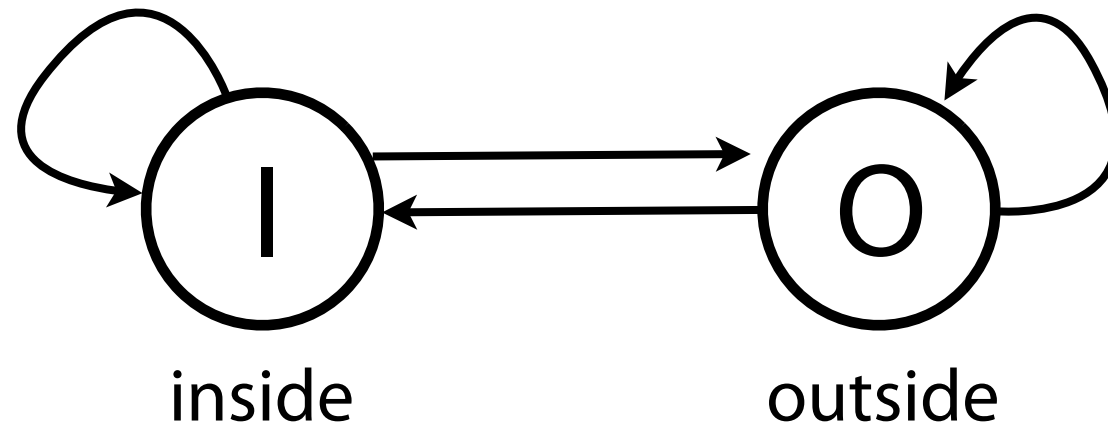
Hidden Markov Model

Idea 1: $Q = \{ \text{inside, outside} \}$, $\Sigma = \{ A, C, G, T \}$



Hidden Markov Model

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A	I	O
I		
O		

Transition matrix

E	A	C	G	T
I				
O				

Emission matrix

Hidden Markov Model

Idea 1: $Q = \{ \text{inside, outside} \}$, $\Sigma = \{ A, C, G, T \}$

Training data consists of bases and associated I/O annotations:

ATCGTCTCGATATAGCGCGGTACGCGATTAAACGATATATATCGGCGTGCCGTACG ...
OOOOOOOOOOOOOOOOOIIIIIIIIIIIOOOOOOOOOOOOOOOOOOIIIIIIIIIIIIIIII ...

A	I	O
I		
O		

Transition matrix

E	A	C	G	T
I				
O				

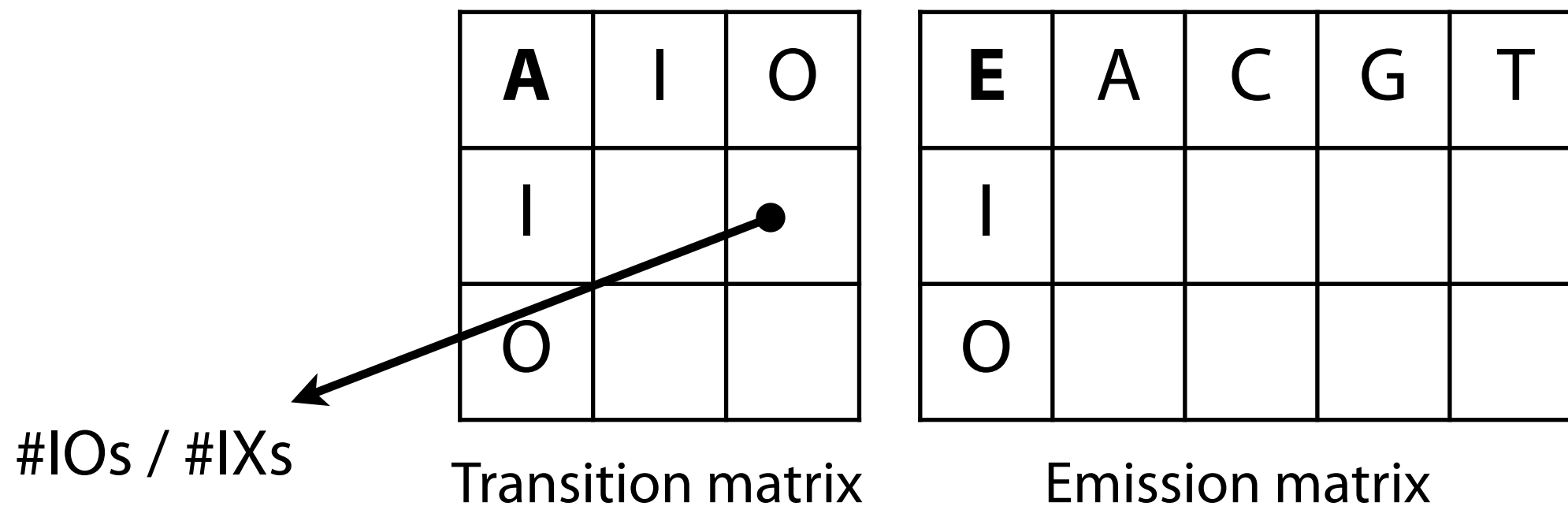
Emission matrix

Hidden Markov Model

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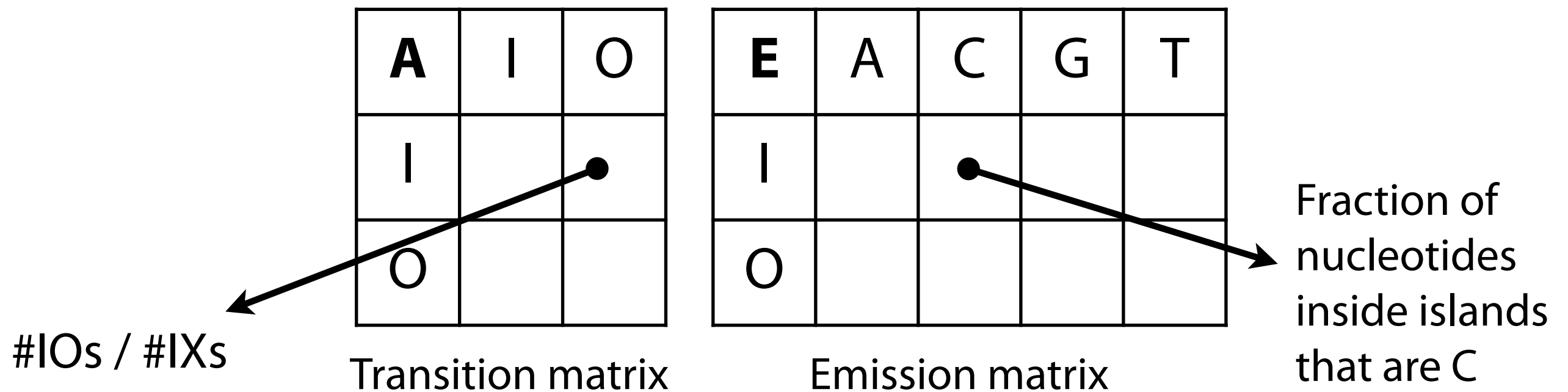


Hidden Markov Model

Idea 1: $Q = \{ \text{inside, outside} \}$, $\Sigma = \{ A, C, G, T \}$

Training data consists of bases and associated I/O annotations:

ATCGTCTCGATATAGCGCGGTACGCGATTAAACGATATATATCGGCGTGCCGTACG ...
00000000000000000000IIIIIIIIIIII00000000000000000000IIIIIIIIIIII ...



Hidden Markov Model

Example 1 using HMM idea 1:

A	I	O	E	A	C	G	T	I
I	0.8	0.2	I	0.1	0.4	0.4	0.1	0.5
O	0.2	0.8	O	0.25	0.25	0.25	0.25	0.5

x: ATATATACGCGCGCGCGCGATATATATATA

p: 00000000IIIIIIIIIIIIII00000000000000

(from Viterbi)

Hidden Markov Model

Example 1 using HMM idea 1:

A	I	O	E	A	C	G	T	I
I	0.8	0.2	I	0.1	0.4	0.4	0.1	0.5
O	0.2	0.8	O	0.25	0.25	0.25	0.25	0.5

x: ATATCGCGCGCGATATATCGCGCGCGATATATAT

Hidden Markov Model

Example 1 using HMM idea 1:

A	I	O	E	A	C	G	T	I
I	0.8	0.2	I	0.1	0.4	0.4	0.1	0.5
O	0.2	0.8	O	0.25	0.25	0.25	0.25	0.5

x: ATATCGCGCGCGATATATCGCGCGCGATATATAT

p: 0000IIIIIIII0000000IIIIIIII0000000000

Hidden Markov Model

Example 1 using HMM idea 1:

A	I	O	E	A	C	G	T	I
I	0.8	0.2	I	0.1	0.4	0.4	0.1	0.5
O	0.2	0.8	O	0.25	0.25	0.25	0.25	0.5

x: ATATATACCCCCCCCCCCCCCATATATATATATA

Hidden Markov Model

Example 1 using HMM idea 1:

A	I	O	E	A	C	G	T	I
I	0.8	0.2	I	0.1	0.4	0.4	0.1	0.5
O	0.2	0.8	O	0.25	0.25	0.25	0.25	0.5

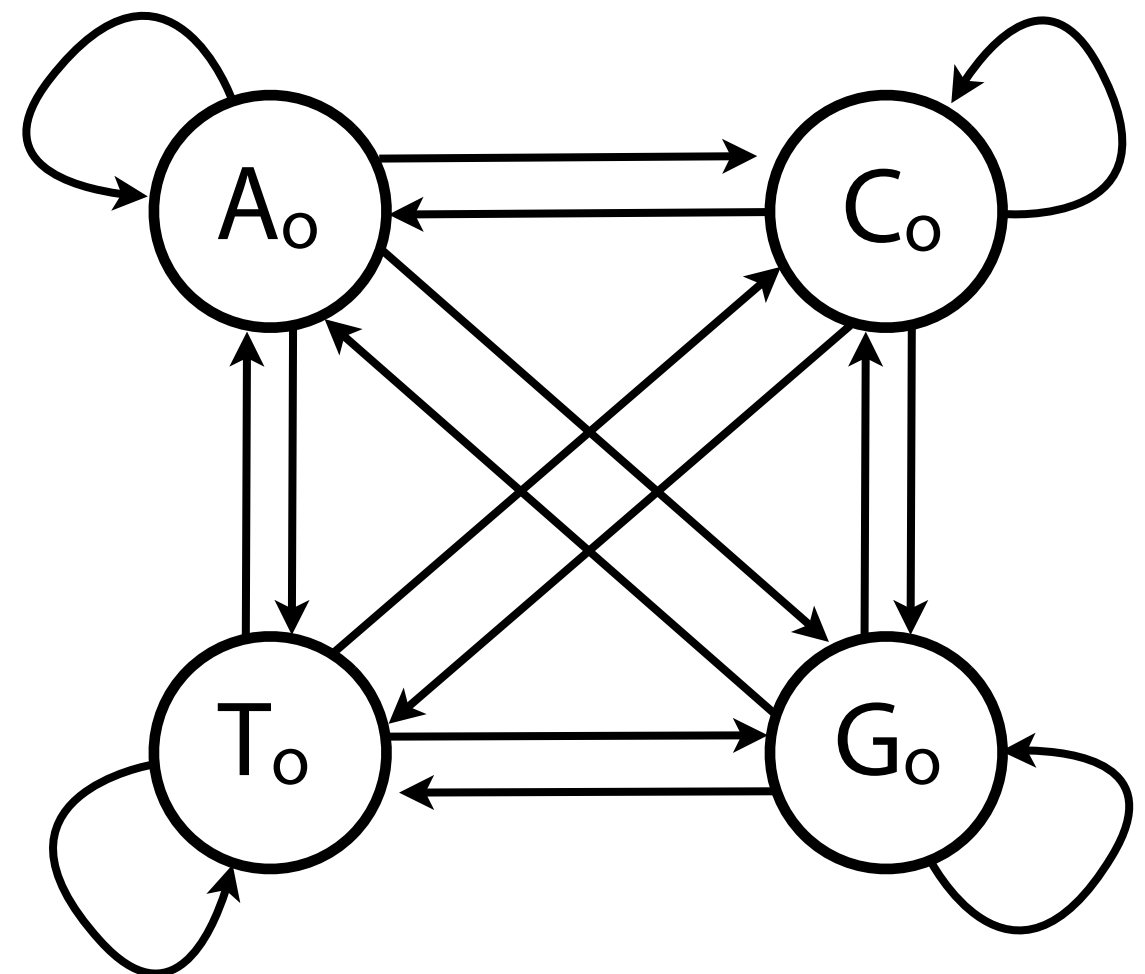
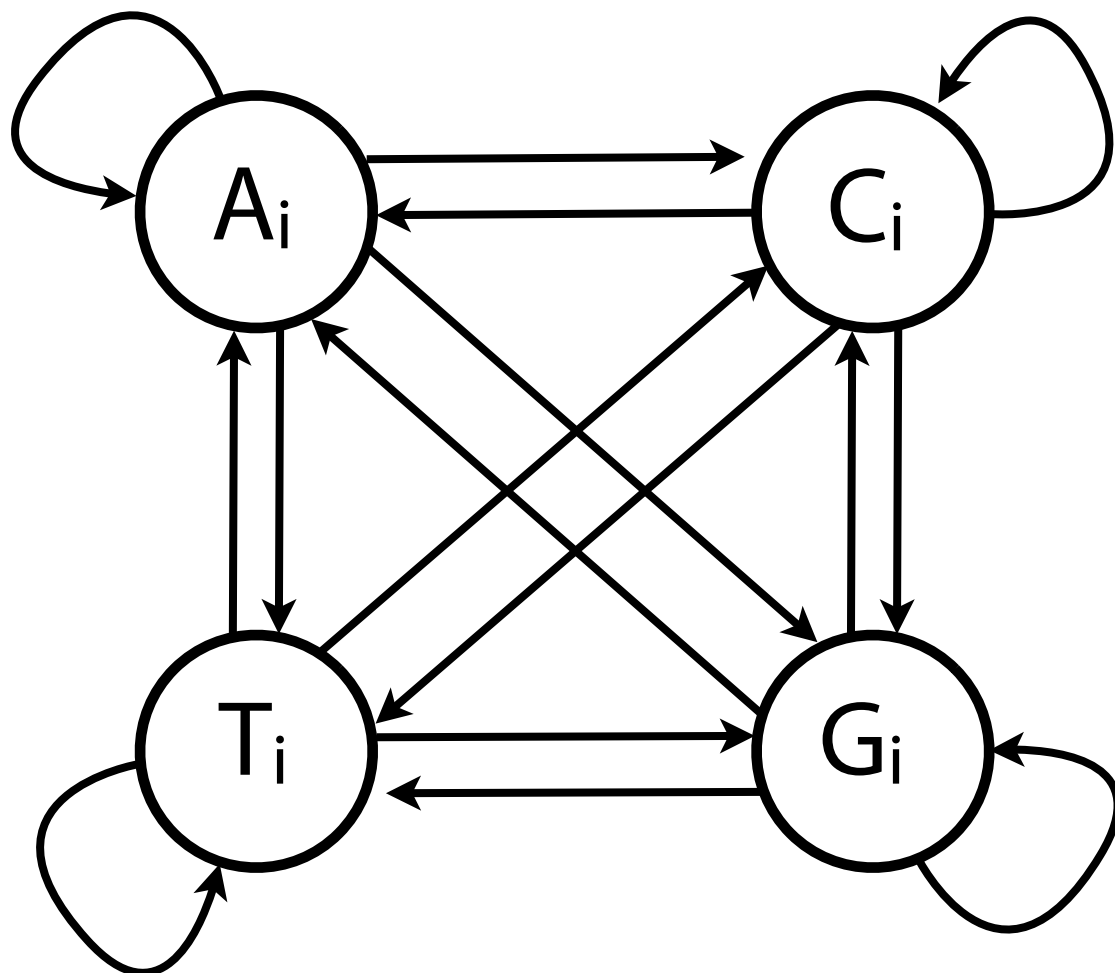
x: ATATATACCCCCCCCCCCCCCATATATATATA

p: 00000000IIIIIIIIIIIIII00000000000000

😬 - clearly not a CpG island

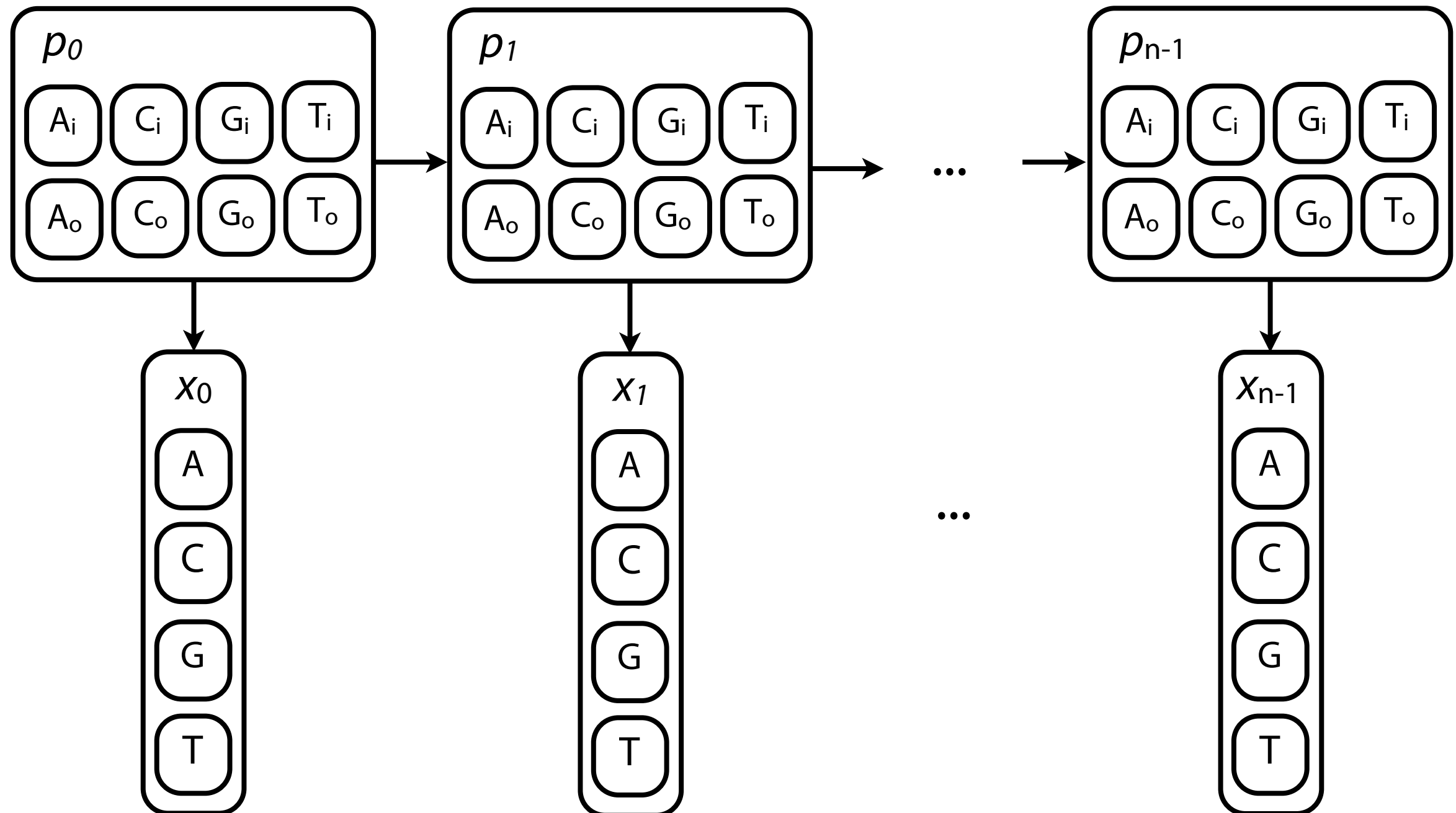
Hidden Markov Model

Recall our 2-chain model from before



Hidden Markov Model

Idea 2: $Q = \{ A_i, C_i, G_i, T_i, A_o, C_o, G_o, T_o \}, \Sigma = \{ A, C, G, T \}$



Hidden Markov Model

Idea 2: $Q = \{ A_i, C_i, G_i, T_i, A_o, C_o, G_o, T_o \}, \Sigma = \{ A, C, G, T \}$

A	A _i	C _i	G _i	T _i	A _o	C _o	G _o	T _o
A _i								
C _i								
G _i								
T _i								
A _o								
C _o								
G _o								
T _o								

Transition matrix

E	A	C	G	T
A _i				
C _i				
G _i				
T _i				
A _o				
C _o				
G _o				
T _o				

Emission matrix

Hidden Markov Model

ATCGTCTCGATATAGCGCGGTACGCGATTAAACGATATATATC GGCGTGCCGTACG
0000000000000000IIIIIIIIIIIO0000000000000000IIIIIIIIIIIII

A	A _i	C _i	G _i	T _i	A _o	C _o	G _o	T _o
A _i								
C _i								
G _i								
T _i								
A _o								
C _o								
G _o								
T _o								

Transition matrix

E	A	C	G	T
A _i				
C _i				
G _i				
T _i				
A _o				
C _o				
G _o				
T _o				

Emission matrix

Hidden Markov Model

ATCGTCTCGATATAGCGCGGTACGCGATTAAACGATATATATCGGGCGTGCCGTACG
00000000000000000000IIIIIIIIIIII00000000000000000000IIIIIIIIIIIIII

A	A _i	C _i	G _i	T _i	A _o	C _o	G _o	T _o
A _i								
C _i								
G _i								
T _i								
A _o								
C _o								
G _o								
T _o								

Estimate $P(C_i | T_i)$ as
$T_i C_i$ s divided by # T_i s

Transition matrix

E	A	C	G	T
A _i				
C _i				
G _i				
T _i				
A _o				
C _o				
G _o				
T _o				

Emission matrix

Hidden Markov Model

ATCGTCTCGATATAGCGCGGTACGCGATTAAACGATATATATCGGGCGTGCCGTACG
00000000000000000000IIIIIIIIIIII00000000000000000000IIIIIIIIIIIIII

A	A _i	C _i	G _i	T _i	A _o	C _o	G _o	T _o
A _i								
C _i								
G _i								
T _i								
A _o								
C _o								
G _o								
T _o								

Estimate $P(C_i | T_i)$ as
$T_i C_i$ s divided by # T_i s

Transition matrix

E	A	C	G	T
A _i	1	0	0	0
C _i	0	1	0	0
G _i	0	0	1	0
T _i	0	0	0	1
A _o	1	0	0	0
C _o	0	1	0	0
G _o	0	0	1	0
T _o	0	0	0	1

Emission matrix

Hidden Markov Model

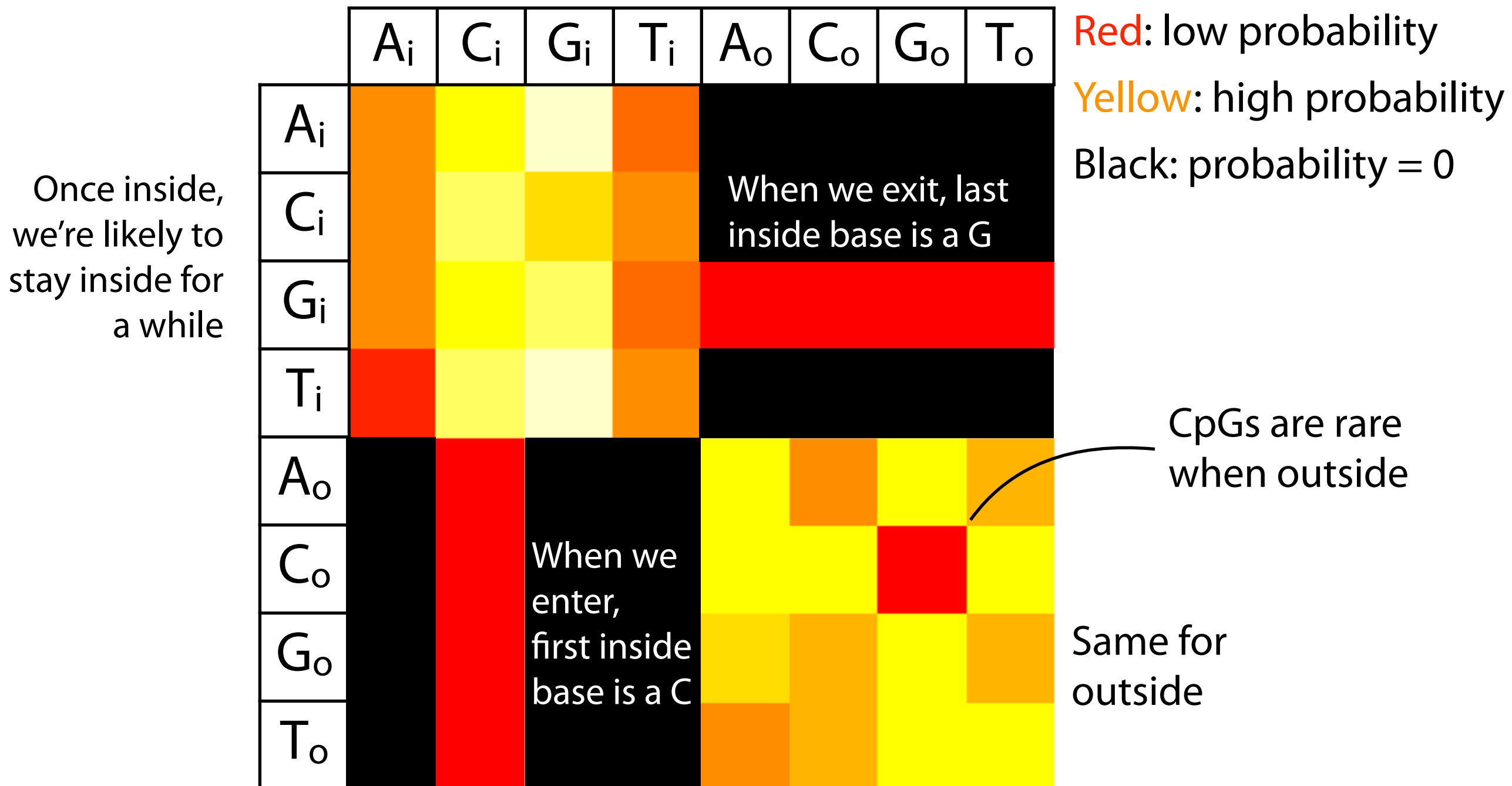
Trained transition matrix:

	A	C	G	T	a	c	g	t
A:	0.185441	0.276405	0.400914	0.137241	0.000000	0.000000	0.000000	0.000000
C:	0.189582	0.359051	0.253240	0.198127	0.000000	0.000000	0.000000	0.000000
G:	0.171904	0.328635	0.354250	0.139893	0.000785	0.001479	0.001857	0.001198
T:	0.094102	0.341636	0.376867	0.187395	0.000000	0.000000	0.000000	0.000000
a:	0.000000	0.000038	0.000000	0.000000	0.294811	0.194641	0.286965	0.223544
c:	0.000000	0.000076	0.000000	0.000000	0.326810	0.294085	0.061724	0.317305
g:	0.000000	0.000057	0.000000	0.000000	0.257128	0.233486	0.294244	0.215085
t:	0.000000	0.000031	0.000000	0.000000	0.179562	0.232469	0.294630	0.293308

Uppercase = inside, lowercase = outside

Hidden Markov Model

Trained transition matrix:



Hidden Markov Model

Viterbi result; lowercase = *outside*, uppercase = *inside*:

[illegible]

Hidden Markov Model

Viterbi result; lowercase = *outside*, uppercase = *inside*:

[illegible]