

See also 18-Apr-2018 and 16-May-2018 at  
<http://phyloseminar.org/recorded.html>

# Likelihood in Phylogenetics

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# Probability density

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The **expected number** of substitutions/site equals the total substitution **rate** multiplied by **time**

$$v = (\text{subst. rate})(\text{time})$$

**Average** of 10 values

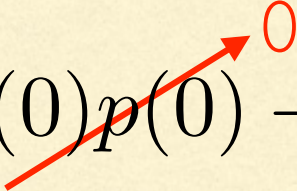
$$\begin{aligned} \text{average} &= \frac{1 + 2 + 4 + 3 + 4 + 3 + 2 + 4 + 3 + 4}{10} \\ &= (0) \left( \frac{0}{10} \right) + (1) \left( \frac{1}{10} \right) + (2) \left( \frac{2}{10} \right) + (3) \left( \frac{3}{10} \right) + (4) \left( \frac{4}{10} \right) + (5) \left( \frac{0}{10} \right) + \dots \end{aligned}$$

**Expected value** of a random variable  $X$

$$E[X] = (0)p(0) + (1)p(1) + (2)p(2) + (3)p(3) + (4)p(4) + \dots$$


Expected value is same as the simple average if the probabilities used are sample relative frequencies

Expected number of substitutions **if only an instant of time is considered** (only 0 or 1 substitutions possible):

$$E[X] = (0)p(0) + (1)p(1) = p(1)$$


Note that **probability of a substitution** equals the **expected number of substitutions** in this case

$$p(\text{subst.}) = \lambda dt$$

substitution rate  instant of time

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# Revisiting your simulations from yesterday

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Yesterday you drew a uniform random variable  $u$  (using your 10-sided die) and transformed it to obtain a time distributed as an Exponential distribution with rate  $\lambda$



A statistician would write

$$t \sim \text{Exponential}(\lambda)$$

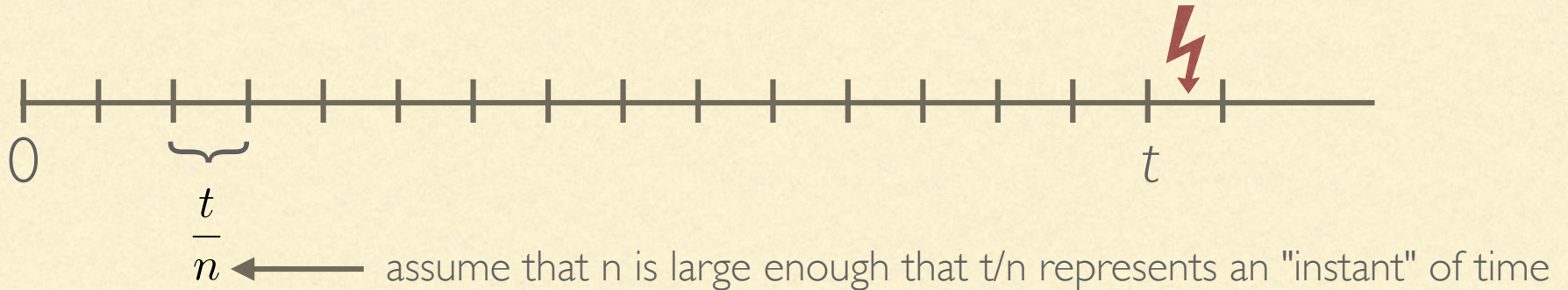
and also would say that the probability **density** of  $t$  is

$$p(t|\lambda) = \lambda e^{-\lambda t}$$

What is a probability density and where did that e come from?

$$p(t|\lambda) = \lambda e^{-\lambda t}$$

Start by imagining that the interval 0 to  $t$  is divided into  $n$  equal segments and our substitution occurs in the very next segment:



The probability that a substitution falls in any segment is

$$\lambda \frac{t}{n}$$

The probability that there is no substitution in a given segment is thus

$$1 - \lambda \frac{t}{n}$$

The probability that *no substitution* occurred in *any* of the  $n$  segments is thus

$$\left(1 - \lambda \frac{t}{n}\right)^n \approx e^{-\lambda t}$$

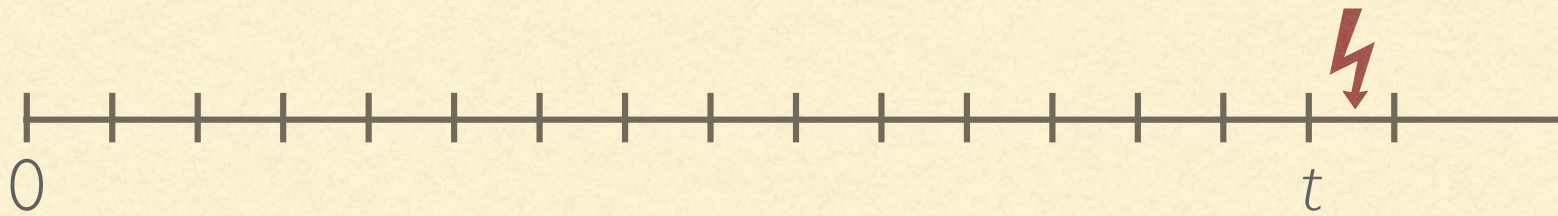
The approximation shown above works well if  $n$  is large (and we can set  $n$  equal to infinity if we like).

More generally,

$$\lim_{n \rightarrow \infty} \left(1 + \frac{a}{n}\right)^n = e^a$$

We're using  $a = -\lambda t$

You can approximate Euler's constant  $e$  by just calculating the above quantity using  $a = 1$  and a large value of  $n$ : for example, for  $n = 1000000$ , the formula gives 2.71828047



We now have the probability that *no substitution* occurred in *any* of the  $n$  segments spanning the interval 0 to  $t$ , so all we need now is the probability that a substitution **did occur** in the very next instant of time, which is just  $\lambda dt$ , where  $dt$  is an infinitesimal time period.

The probability of seeing any substitution at exactly time  $t$  is

$$e^{-\lambda t} (\lambda dt)$$

This can be factored into a probability **density** and  $dt$ :

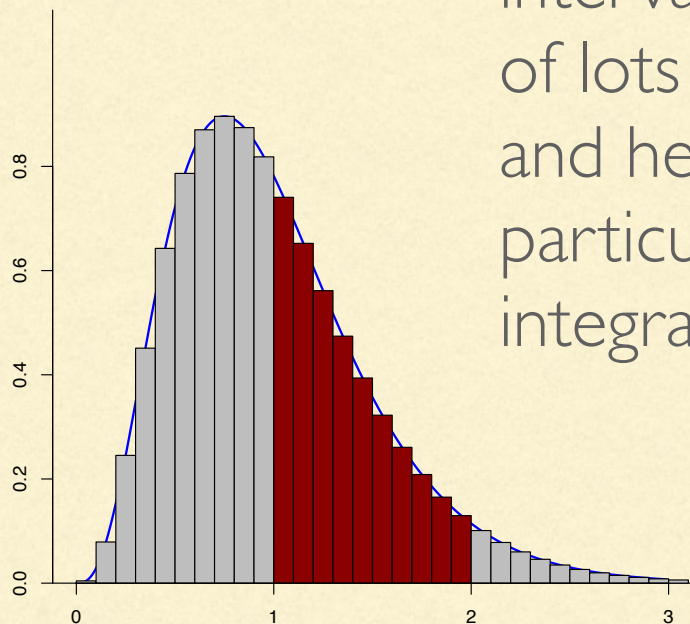
$$(\lambda e^{-\lambda t}) dt$$



A probability density allows you to calculate a probability if multiplied by a time interval, much like the density of gold allows you to compute the mass of gold in a specified volume.

A big difference is that gold has the same density throughout, whereas the probability density is only valid for one value of  $t$ ; the density continuously changes with  $t$ .

To compute the probability that  $t$  is in the interval 1 to 2 we need to add up the volumes of lots of rectangles each of which has width  $dt$  and height equal to the density function at a particular value of  $t$  (i.e. we need to do integration):



$$\Pr (1 \leq t \leq 2) = \int_0^1 \lambda e^{-\lambda t} dt$$

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# Probability

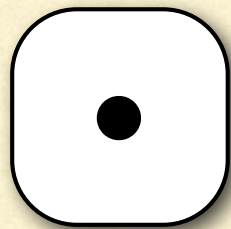
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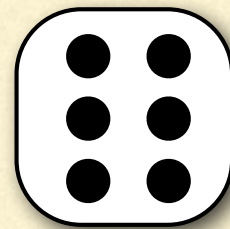
# Probabilities: the AND rule

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Rolling 2 dice, what is the probability of seeing (simultaneously) a 1 on the first die and a 6 on the second die?



AND



$(1/6)$

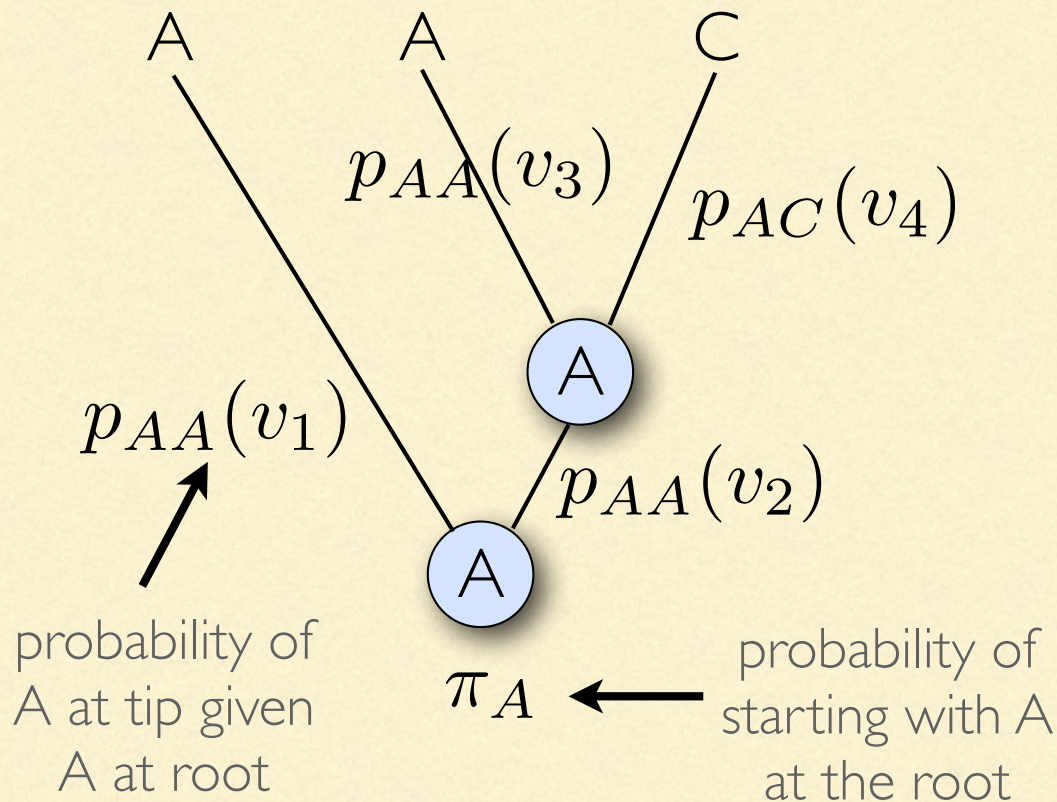
$\times$

$(1/6)$

$=$

$1/36$

# AND rule in phylogenetics



One use of the AND rule in phylogenetics is to combine probabilities associated with individual branches to produce the overall probability of the data for one site.

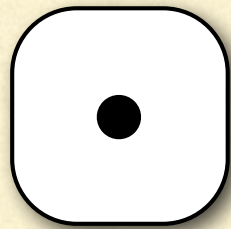
$$\Pr(A, A, C, A, A) = \pi_A p_{AA}(v_1) p_{AA}(v_2) p_{AA}(v_3) p_{AC}(v_4)$$

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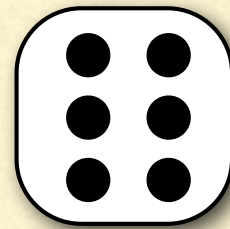
# Probabilities: the OR rule

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Rolling 1 die, what is the probability of seeing either a 1 or a 6?



OR



$(1/6)$

+

$(1/6)$

=

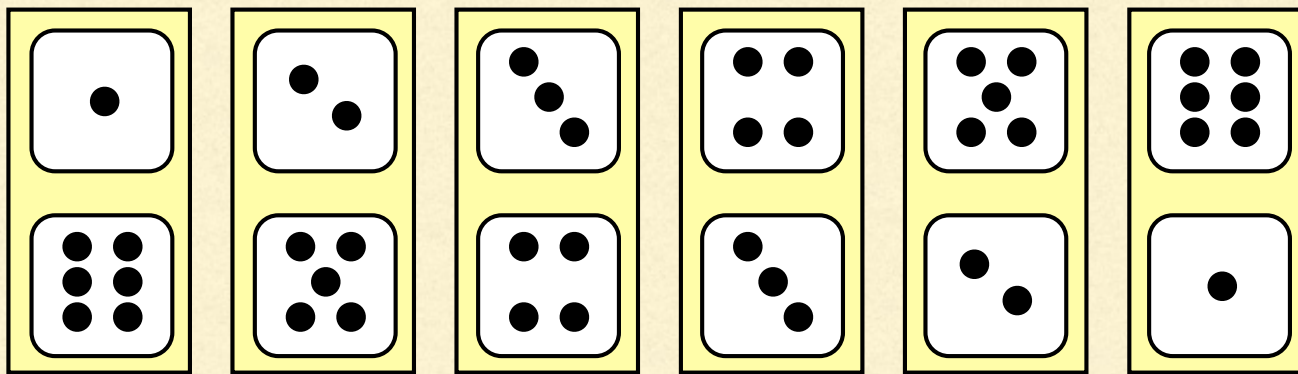
$1/3$

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# Combining AND and OR

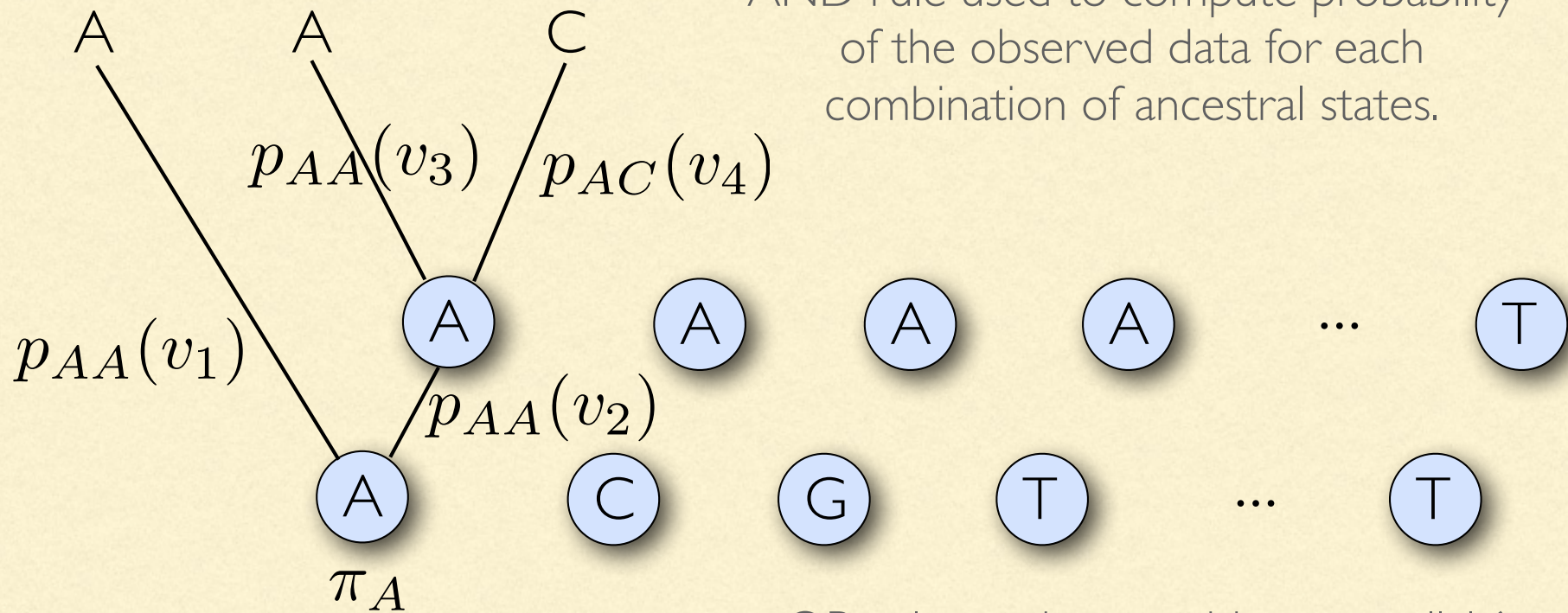
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What is the probability that the sum of two dice is 7?



$$(1/36) + (1/36) + (1/36) + (1/36) + (1/36) + (1/36) = 1/6$$

# Using both AND and OR in phylogenetics



AND rule used to compute probability of the observed data for each combination of ancestral states.

OR rule used to combine over all 16 combinations of ancestral states.

$$\Pr(\mathbf{A,A,C}) = \Pr(\mathbf{A,A,C},A,A) + \Pr(\mathbf{A,A,C},A,C) + \dots + \Pr(\mathbf{A,A,C},T,T)$$

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# Independence

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$$\Pr(A, B) = \Pr(A) \Pr(B)$$

Probability of flipping a coin twice and getting heads both times:

$$\Pr(H, H) = \Pr(H) \Pr(H)$$



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# Non-independence

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$$\Pr(A, B) = \Pr(A) \Pr(B|A)$$

↑ joint probability of A and B                                  ↑ conditional probability of B given A

$$\Pr(\text{walk to work}|\text{sunny}) = 0.99$$
$$\Pr(\text{walk to work}|\text{raining}) = 0.50$$

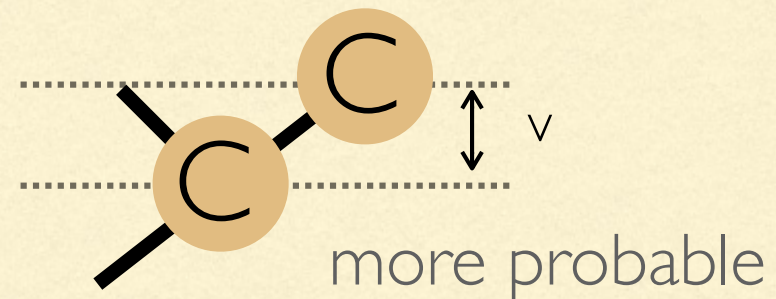
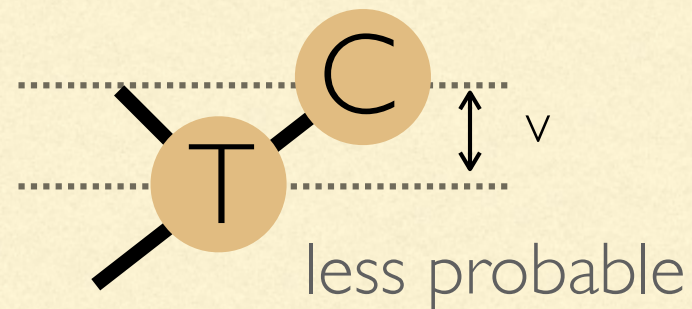
# Non-independence in phylogenies

Normally, for a given rate of substitution and time, the probability of the end state is *dependent* on the starting state

$$p(C|C, v) > p(C|T, v)$$

$$p_{CC}(v) > p_{TC}(v)$$

common notation for transition probabilities

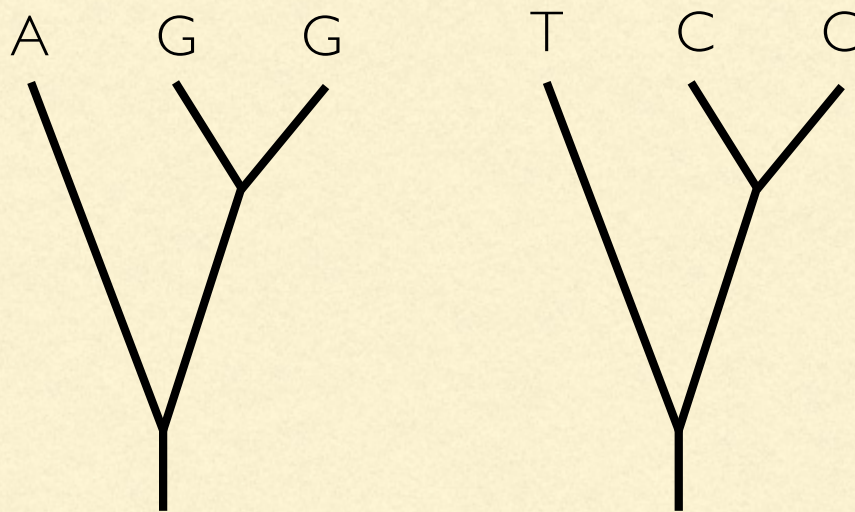


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# Conditional Independence

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$$\Pr(A,B|C) = \Pr(A|C) \Pr(B|C)$$



$$\Pr(AGG,TCC|\text{tree}) = \Pr(AGG|\text{tree}) \Pr(TCC|\text{tree})$$

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# Back to your simulations...

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You first chose a waiting time  $t$  until the next substitution, and then you used your dice again to choose which nucleotide was actually substituted (G in the case shown below)



To make this choice, we need to calculate the conditional probability of an  $A \rightarrow G$  substitution **given** that a substitution occurred. We can manipulate a formula shown previously to get the conditional probability we want:

$$\Pr(A, B) = \Pr(A) \Pr(B|A) \longrightarrow \Pr(B|A) = \frac{\Pr(A, B)}{\Pr(A)}$$

Replacing A with "substitution" and B with "A → G"

$$\Pr(A \rightarrow G | \text{substitution}) = \frac{\Pr(A \rightarrow G, \text{substitution})}{\Pr(\text{substitution})}$$

Note that the joint probability  $\Pr(A \rightarrow G, \text{substitution})$  equals  $\Pr(A \rightarrow G)$  because a substitution must have occurred if an A changed to a G.

$$= \frac{r_{AG} dt}{r_{AC} dt + r_{AG} dt + r_{AT} dt}$$

$$= \frac{r_{AG}}{r_{AC} + r_{AG} + r_{AT}}$$

←——— lambda ———→

The probability of any substitution at exactly time  $t$  is  $\lambda dt$ , but the rate of any substitution ( $\lambda$ ) is just the sum of the rates of all possible substitutions ( $r_{AC} + r_{AG} + r_{AT}$ )

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# Likelihood

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# Why do we need the term **likelihood**?

Outcome	Fair coin model	Two-heads model
H	0.5	1
T	0.5	0
	1	1

Likelihoods of models given one particular data outcome are not expected to sum to 1.0

Probabilities of data outcomes given one particular model sum to 1.0

**Probability** of the **data**  
given the model  
**Likelihood** of the **model**  
given the data

# Likelihood of a single vertex

First 32 nucleotides of the  $\psi\eta$ -globin gene of gorilla:

● **GAAGTCCTTGAGAAATAAACTGCACACTGG**

$$L = \Pr(G) \Pr(A) \Pr(A) \Pr(G) \Pr(T) \cdots \Pr(G)$$

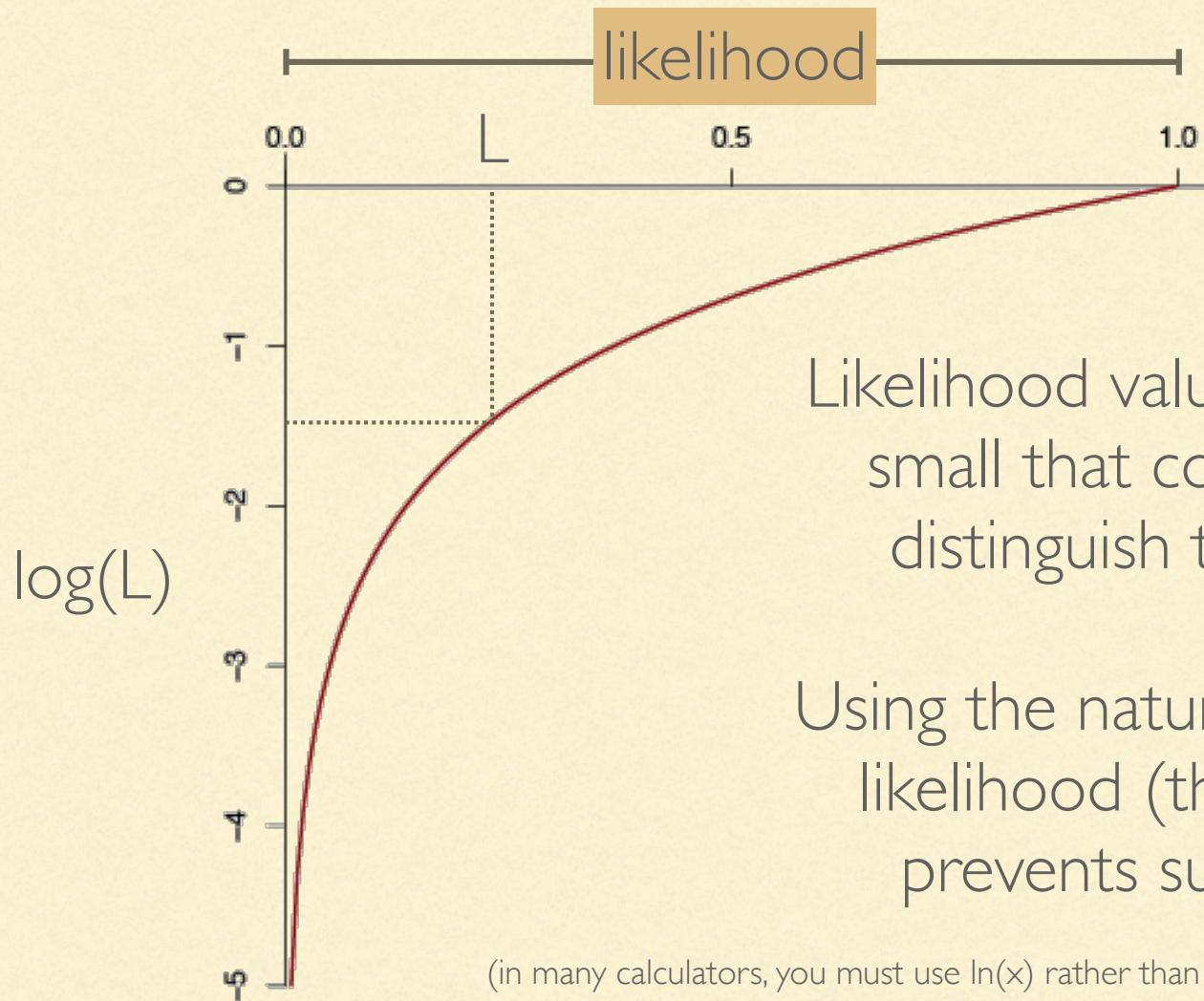
$$L = \pi_G \pi_A \pi_A \pi_G \pi_T \cdots \pi_G$$

$$L = \pi_A^{12} \pi_C^7 \pi_G^7 \pi_T^6$$

$$\log L = 12 \log(\pi_A) + 7 \log(\pi_C) + 7 \log(\pi_G) + 6 \log(\pi_T)$$



# Natural logarithm



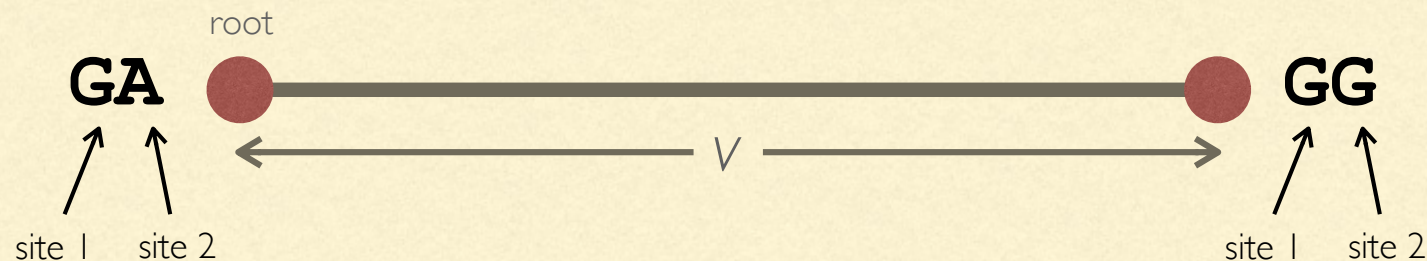
Likelihood values can become so small that computers cannot distinguish them from zero.

Using the natural logarithm of the likelihood (the log-likelihood) prevents such "underflow"

(in many calculators, you must use  $\ln(x)$  rather than  $\log(x)$  to take the natural log of the value  $x$ )

# Likelihood of a single-edge tree

Two nodes have sequence data (but only for two sites)



$$L = \left[ \left( \frac{1}{4} \right) \left( \frac{1}{4} + \frac{3}{4} e^{-4v/3} \right) \right] \left[ \left( \frac{1}{4} \right) \left( \frac{1}{4} - \frac{1}{4} e^{-4v/3} \right) \right]$$

$\leftarrow$  site 1  $\rightarrow$        $\leftarrow$  site 2  $\rightarrow$

Each **site likelihood** is the probability of the **starting state** at the root ( $1/4$ ) times the **transition probability** (probability of the end state given the starting state)

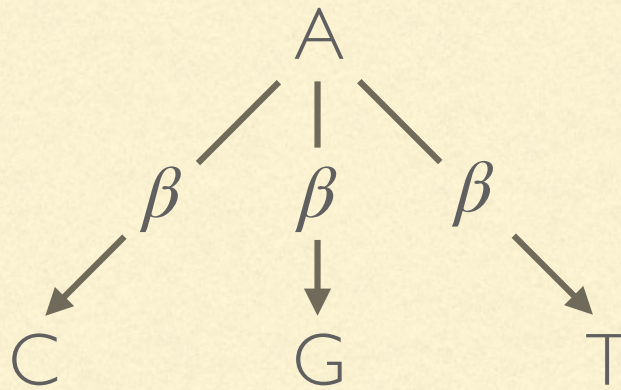
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# What is the edge length $v$ ?

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expected  
**number** of  
substitutions  
per site = substitution  
**rate** per site  $\times$  **time**

$$v = 3\beta \times t$$



3 possible substitutions, each of which happens with rate  $\beta$

# Jukes and Cantor (1969)

JC69 model

to:

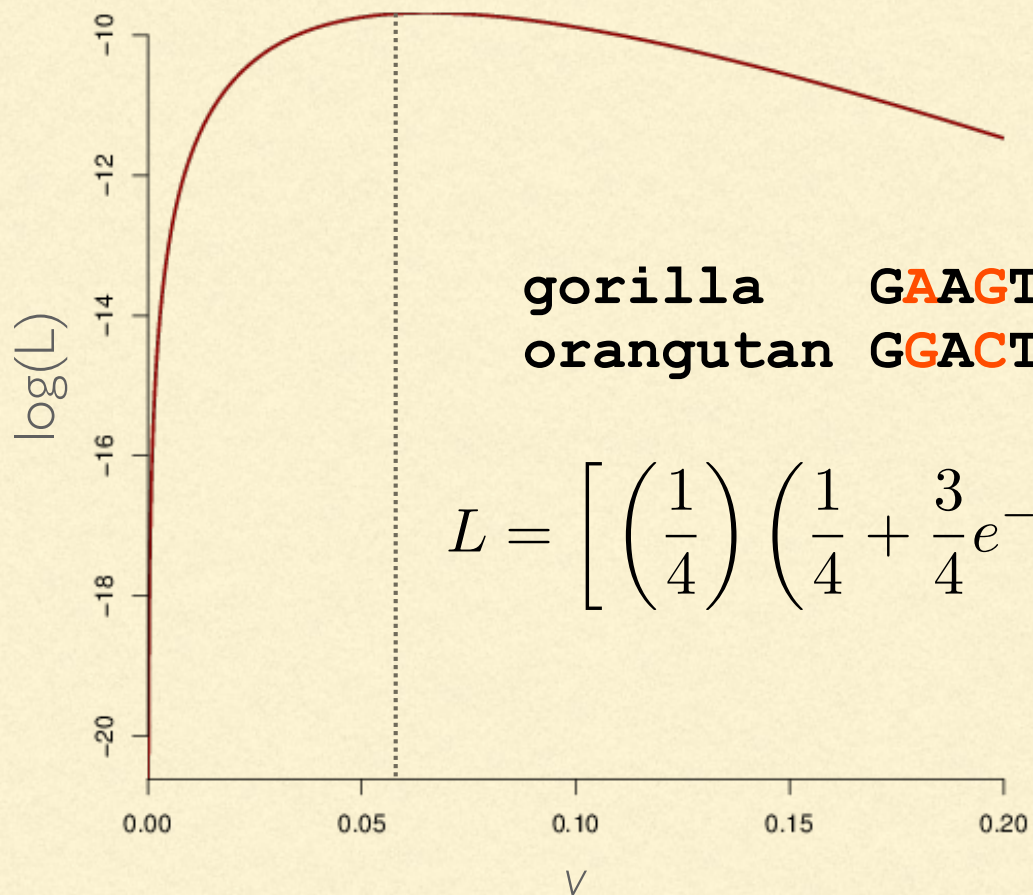
Parameters:  $\beta$

from:

	A	C	G	T
A	$-3\beta$	$\beta$	$\beta$	$\beta$
C	$\beta$	$-3\beta$	$\beta$	$\beta$
G	$\beta$	$\beta$	$-3\beta$	$\beta$
T	$\beta$	$\beta$	$\beta$	$-3\beta$

# Maximum likelihood estimation

0.065 is the maximum likelihood estimate (MLE) of  $\nu$



gorilla **GAAG**TCCTTGAGAAATAAACTGCACACACTGG  
orangutan **GGAC**TCCTTGAGAAATAAACTGCACACACTGG

$$L = \left[ \left( \frac{1}{4} \right) \left( \frac{1}{4} + \frac{3}{4} e^{-4\nu/3} \right) \right]^{30} \left[ \left( \frac{1}{4} \right) \left( \frac{1}{4} - \frac{1}{4} e^{-4\nu/3} \right) \right]^2$$

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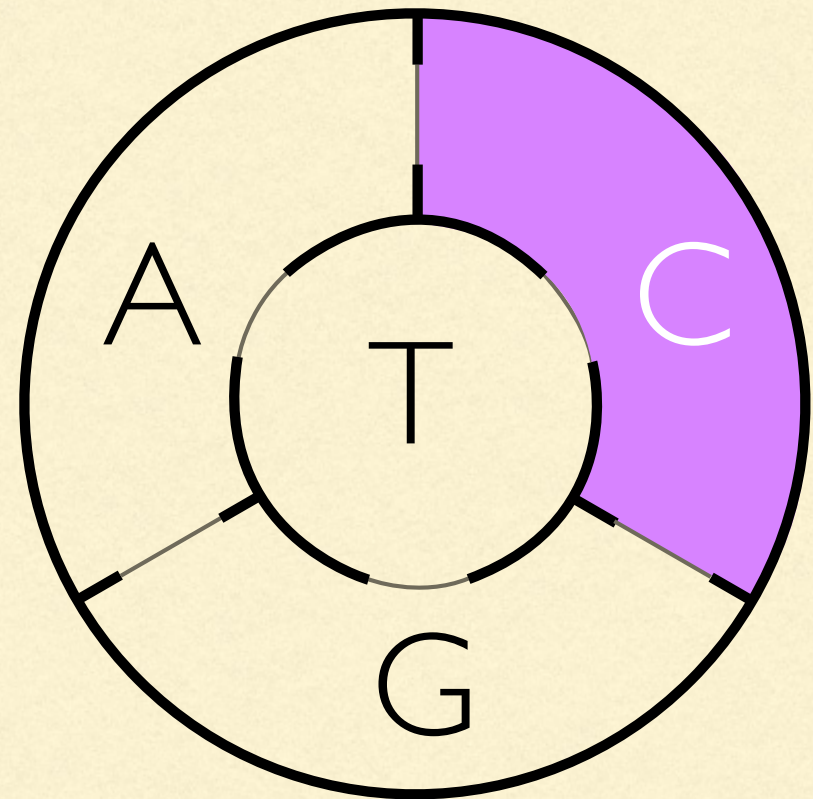
# Equilibrium Frequencies

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Imagine a bottle of perfume has been spilled in room C.

The doors to the other rooms are closed, so the perfume has, thus far, not been able to spread.

What would happen if we opened all the doors?



Architect: Joe Bielawski

# Equilibrium Frequencies

At the **instant the doors open**, perfume molecules...

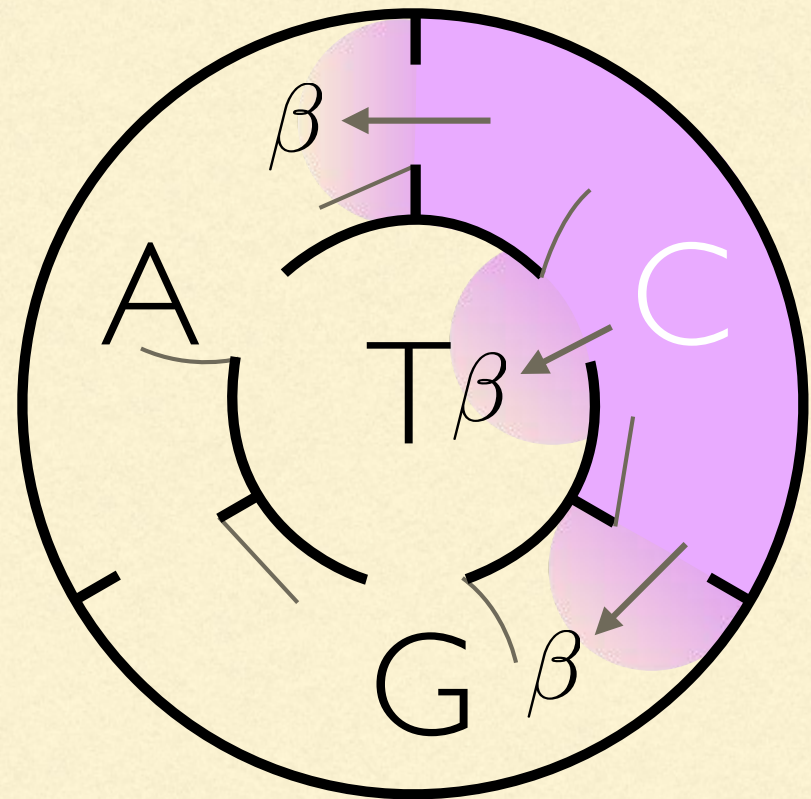
enter room A at rate  $\beta$

enter room T at rate  $\beta$

enter room G at rate  $\beta$

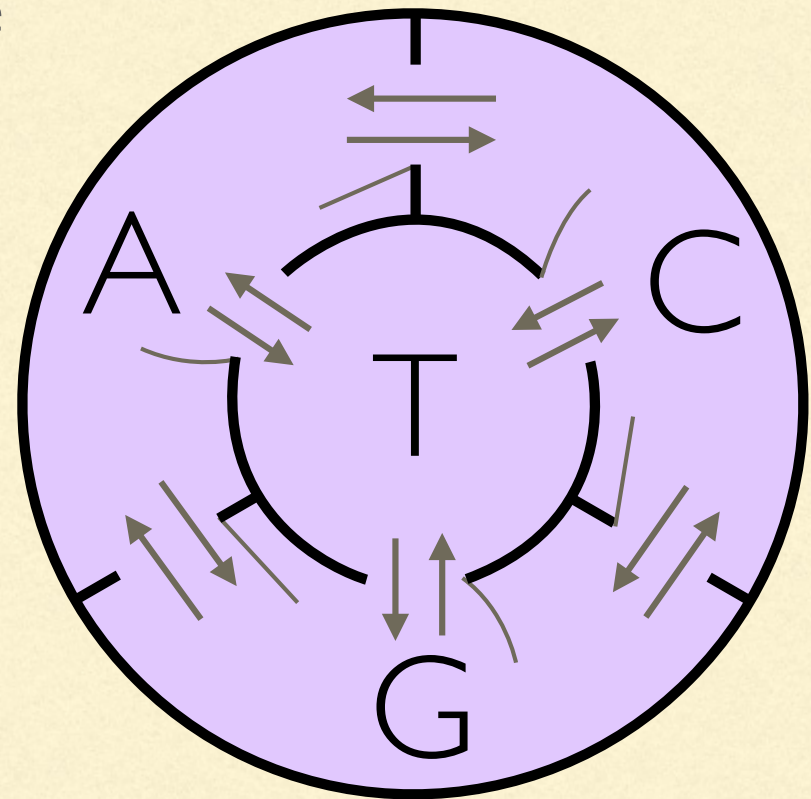
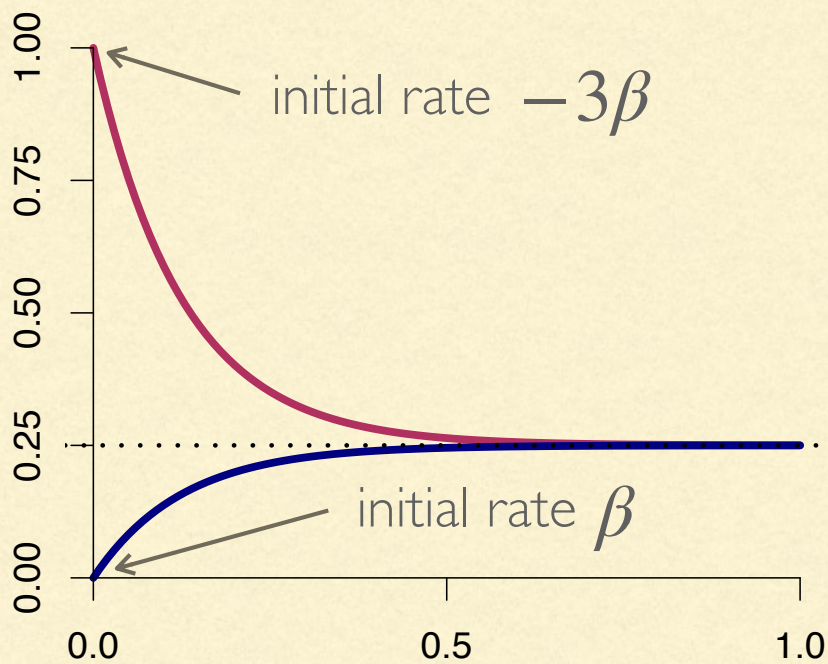
enter room C at rate  $-3\beta$

(you could also say they *leave* C at rate  $3\beta$ )



# Equilibrium Frequencies

At **equilibrium**, the relative concentration of perfume is **equal** in all rooms



$$\pi_A = \pi_C = \pi_G = \pi_T = \frac{1}{4}$$



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# Transition probability demo

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<https://plewis.github.io/applets/jc-transition-probabilities/>

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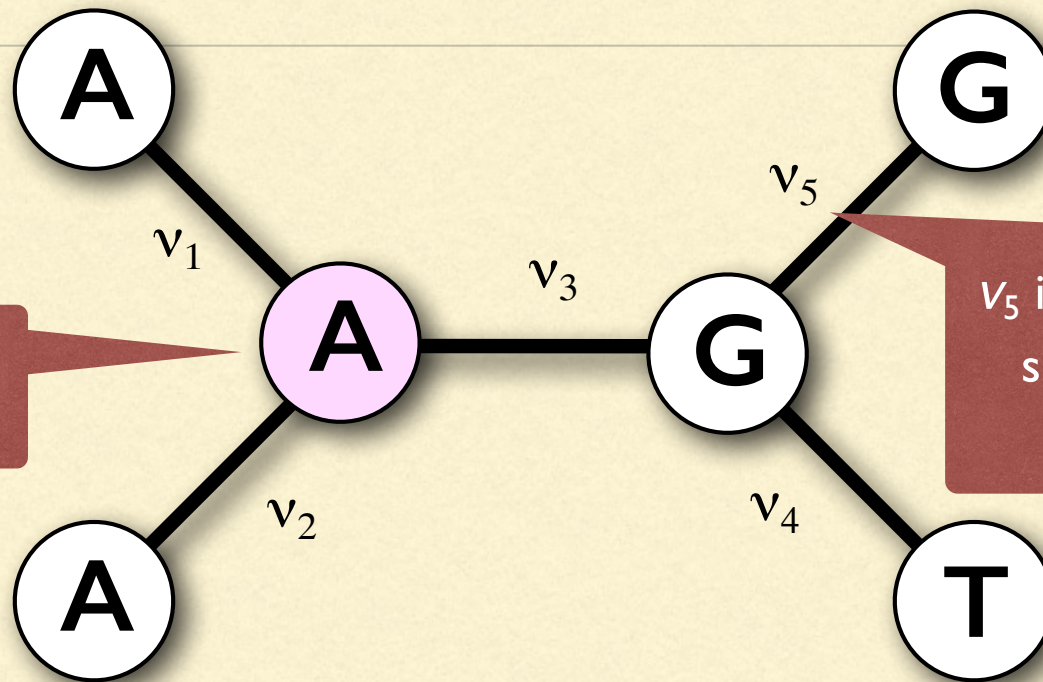
# Sequence data for four taxa

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one site

Sphagnum	GGCAGCATTTCGAATGACTCCTCAACCTGGAGT	<b>A</b>	CACCCG...
Asplenium	GGCAGCTTTCCGGATGACCCACAACCCGGAGT	<b>A</b>	CAGCTG...
Picea	GGCAGCATTCCGAGTAACTCCTCAACCAGGGGT	<b>G</b>	CGCCG...
Avena	GGCAGCATTCCGAGTAACTCCTCAACCTGGGGT	<b>T</b>	CGCCG...

# Likelihood for tree (one site)



$$L = \frac{1}{4} \left[ \frac{1}{4} + \frac{3}{4} e^{-4\nu_1/3} \right] \left[ \frac{1}{4} + \frac{3}{4} e^{-4\nu_2/3} \right] \left[ \frac{1}{4} - \frac{1}{4} e^{-4\nu_3/3} \right] \left[ \frac{1}{4} - \frac{1}{4} e^{-4\nu_4/3} \right] \left[ \frac{1}{4} + \frac{3}{4} e^{-4\nu_5/3} \right]$$

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# Total likelihood

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$$L = L_1 L_2 \cdots L_n$$

↑      ↑      ↑  
site 1   site 2      site n


$$\log L = \log L_1 + \log L_2 + \cdots + \log L_n$$

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# Jukes and Cantor (1969)

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JC69 model

to:

Parameters:  $\beta$

from:

$$\begin{array}{c} \text{A} \\ \text{C} \\ \text{G} \\ \text{T} \end{array} \begin{bmatrix} \text{A} & \text{C} & \text{G} & \text{T} \\ -3\beta & \beta & \beta & \beta \\ \beta & -3\beta & \beta & \beta \\ \beta & \beta & -3\beta & \beta \\ \beta & \beta & \beta & -3\beta \end{bmatrix}$$

# Kimura (1980)

K80 (or K2P) model

Parameters:  $\alpha, \beta$

$$\begin{array}{c} \text{A} \\ \text{C} \\ \text{G} \\ \text{T} \end{array} \begin{bmatrix} \text{A} & \text{C} & \text{G} & \text{T} \\ -\alpha - 2\beta & \beta & \alpha & \beta \\ \beta & -\alpha - 2\beta & \beta & \alpha \\ \alpha & \beta & -\alpha - 2\beta & \beta \\ \beta & \alpha & \beta & -\alpha - 2\beta \end{bmatrix}$$

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# Kimura (1980)

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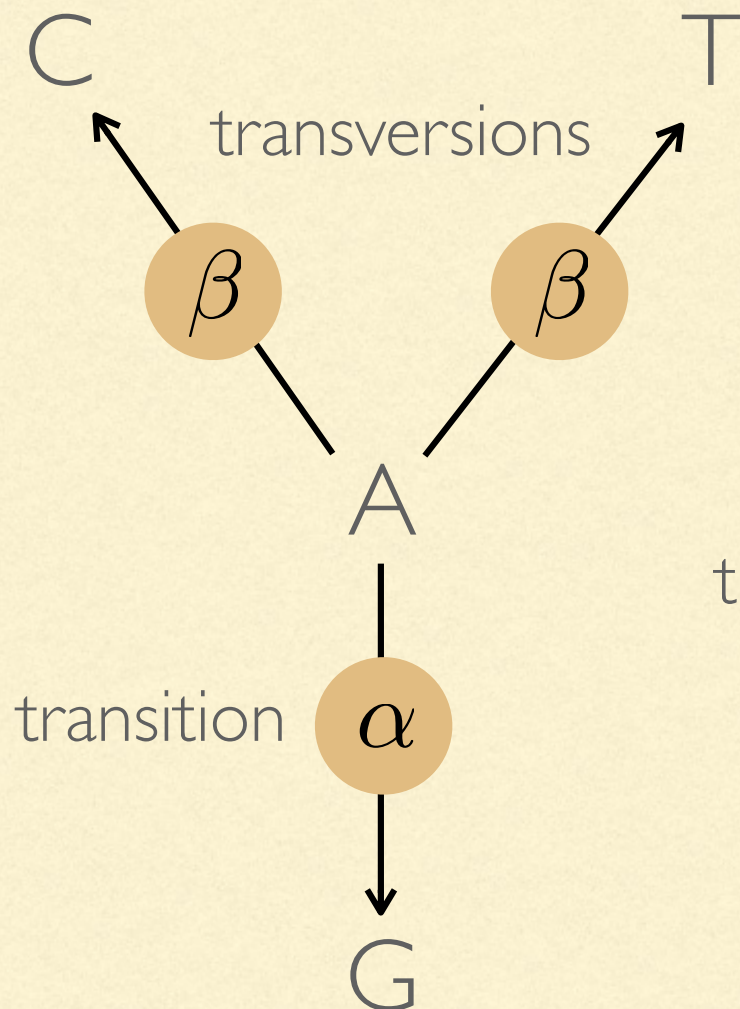
K80 (or K2P) model

$$\kappa = \alpha/\beta$$

Parameters:  $\kappa, \beta$

	A	C	G	T
A	$-\beta(\kappa + 2)$	$\beta$	$\kappa\beta$	$\beta$
C	$\beta$	$-\beta(\kappa + 2)$	$\beta$	$\kappa\beta$
G	$\kappa\beta$	$\beta$	$-\beta(\kappa + 2)$	$\beta$
T	$\beta$	$\kappa\beta$	$\beta$	$-\beta(\kappa + 2)$

# Transition-transversion (rate) ratio



transition rate =  $\alpha$

transversion rate =  $\beta$

assume  $\alpha = \beta$

transition-transversion rate ratio = 1.0

transition-transversion ratio = 0.5



# Felsenstein (1981)

F81 model

Parameters:  $\mu, \pi_A, \pi_C, \pi_G$

	A	C	G	T
A	$-\mu(1 - \pi_A)$	$\pi_C\mu$	$\pi_G\mu$	$\pi_T\mu$
C	$\pi_A\mu$	$-\mu(1 - \pi_C)$	$\pi_G\mu$	$\pi_T\mu$
G	$\pi_A\mu$	$\pi_C\mu$	$-\mu(1 - \pi_G)$	$\pi_T\mu$
T	$\pi_A\mu$	$\pi_C\mu$	$\pi_G\mu$	$-\mu(1 - \pi_T)$

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JC69 is a special case of F8 I

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	A	C	G	T		A	C	G	T
A	$-\frac{3}{4}\mu$	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$	]	$-3\beta$	$\beta$	$\beta$	$\beta$
C	$\frac{1}{4}\mu$	$-\frac{3}{4}\mu$	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$		$\beta$	$-3\beta$	$\beta$	$\beta$
G	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$	$-\frac{3}{4}\mu$	$\frac{1}{4}\mu$		$\beta$	$\beta$	$-3\beta$	$\beta$
T	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$	$\frac{1}{4}\mu$	$-\frac{3}{4}\mu$		$\beta$	$\beta$	$\beta$	$-3\beta$

$$\beta = \frac{1}{4}\mu$$

# Hasegawa, Kishino, and Yano (1985)

HKY85 model

Parameters:  $\mu, \kappa, \pi_A, \pi_C, \pi_G$

one parameter in each model is associated with the length of an edge

these are global parameters (apply to all edge lengths)

	A	C	G	T
A	$-\mu(\pi_C + \pi_G\kappa + \pi_T)$	$\pi_C\mu$	$\pi_G\mu\kappa$	$\pi_T\mu$
C	$\pi_A\mu$	$-\mu(\pi_A + \pi_G + \pi_T\kappa)$	$\pi_G\mu$	$\pi_T\mu\kappa$
G	$\pi_A\mu\kappa$	$\pi_C\mu$	$-\mu(\pi_A\kappa + \pi_C + \pi_T)$	$\pi_T\mu$
T	$\pi_A\mu$	$\pi_C\mu\kappa$	$\pi_G\mu$	$-\mu(\pi_A + \pi_C\kappa + \pi_G)$

# Tavaré (1986)

GTR model

Parameters: ?

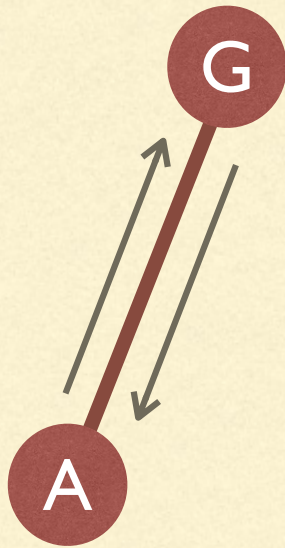
$$\begin{array}{c} \text{A} \\ \text{C} \\ \text{G} \\ \text{T} \end{array} \begin{bmatrix} \text{A} & \text{C} & \text{G} & \text{T} \\ \text{—} & \pi_{C\mu} a & \pi_{G\mu} b & \pi_{T\mu} c \\ \pi_{A\mu} a & \text{—} & \pi_{G\mu} d & \pi_{T\mu} e \\ \pi_{A\mu} b & \pi_{C\mu} d & \text{—} & \pi_{T\mu} f \\ \pi_{A\mu} c & \pi_{C\mu} e & \pi_{G\mu} f & \text{—} \end{bmatrix}$$

exchangeability  
parameters are  
circled

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# Time reversibility

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Time reversibility means...

$$\Pr(A) \Pr(G|A, v) = \Pr(G) \Pr(A|G, v)$$

Time reversibility allows any point on the tree to serve as the root, and thus has some practical advantages, but time reversibility is not a requirement for substitution models used in phylogenetics

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# Rate heterogeneity

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# Green plant rbcL gene

First 88 amino acids (translation is for *Zea mays*)

M--S--P--Q--T--E--T--K--A--S--V--G--F--K--A--G--V--K--D--Y--K--L--T--Y--Y--T--P--E--Y--E--T--K--D--T--D--I--L--A--A--F--R--V--T--P--		
Chara	(green alga; land plant lineage)	AAAGATTACAGATTAACCTTACTATACTCCTGAGTATAAACTAAAGATACTGACATTTTAGCTGCATTTTCGTGTAACCTCCA
Chlorella	(green alga)	....C...C.T.....T..CC..C.A....C.....T..C.T..A..G..C...A.G.....T
Volvox	(green alga)	.....TC.T....A....C..A....C...GT.GTA....C.....C.....A.....A.G.....
Conocephalum	(liverwort)	.....TC.....T.....G..T...G.....G..T.....A.....A.AA.G.....T
Bazzania	(moss)	.....T.....C..T....G....A...G.G..C....G..A..T....G..A.....A.G....C
Anthoceros	(hornwort)	.....T.....CC.T....C....T..CG.G..C..G.....T....G..A..G.C.T.AA.G.....T
Osmunda	(fern)	.....TC...G...C.....C..T...G.G.C.C.G.....T....G..A...C...AA.G....C
Lycopodium	(club "moss")	.GG.....C.T.C.....T....G..C....A..C..T...C.G..A.....AA.G....T
Ginkgo	(gymnosperm; Ginkgo biloba)	.....G.....T.....A...C....C.....T..C..G..A....C..A.....T
Picea	(gymnosperm; spruce)	.....T.....A...C.G..C.....G..T....G..A....C..A.....T
Iris	(flowering plant)	.....G.....T.....T..CG...C.....T..C..G..A....C..A.....T
Asplenium	(fern; spleenwort)	.....TC..C.G....T..C..C..C..A..C..G..C.....C..T..C..G..A..T..C..GA.G..C...
Nicotiana	(flowering plant; tobacco)	....G...A..G....T.....CC...C..G.....T..A..G..A....C..A.....T

Q--L--G--V--P--P--E--E--A--G--A--A--V--A--A--E--S--S--T--G--T--W--T--T--V--W--T--D--G--L--T--S--L--D--R--Y--K--G--R--C--Y--H--I--E--		
CAACCTGGCGTTCCACCTGAAGAAGCAGGGGCTGCAGTAGCTGCAGAATCTTCTACTGGTACATGGACTACTGTTTGGACTGACGGATTAAGTATTGGACCGATACAAAGGAAGATGCTACGATATGAA		
....A..T.....A.....G..T..G.....A.....A..A.....T....G....A.....T..T.....A.....T.....TC..T..T..T..C..C..G		
....A..T.....TGT..T....T..T....T....A..A..A....T....A....A.....T..T....A...C.T....T.....TC..T..T..T..C..C..G		
..G.....G..A...G..A.....A..A....T....T.....A.....T..TC..T...ACC..T..T..T..T....TC.....T.G.....C		
....G..A..A.....A..G.....T....A..C....G....C..G.....C..T..GC..T..A...C..C..T..T.....TC.....T..C..C...		
T...A..G..G.....A..C.....T....A.....C..T..C..T..C..CC..T....T.....TC.....C.....		
....C..A..A..GG...G....T..A.....G.....A....G....C....A...G..T...C..T..C..C..T..T..T..G..TC.....		
....T...A..A....C..G....G..A..C.....T.....C.....C..T..C..T..C..C..C..T..C.....TC.G....T..A.....		
....A..G....G....G..A....C.....C.....C..T..C..T..C..C..T..T..T..G.....T..C..C..G		
....A..G..G..C..G....G..A..A....T....C..C.....C.....C..T..C..T..C..C..T..T..T..G..GC.....T..C..C..G		
....C..A...TG.....G....C..G....C.....A..A..G.....T..C..T..C..C..T..T..T.....C.....C..C..G		
....C..A..A...G.....C..A.....G..C....A.....C...G....A...G..G..C..CC..T....T....G..CC.....C..G		
....A.....C..G.....C.....A.....A...C..T..C..T..C..CC..T..T..T.....GC.....CGC...C..G		

All 4 bases are observed at some sites...

...while at other sites, only 1 base is observed

# Site-specific rates

Each defined subset (e.g. 1st+2nd pos. versus 3rd pos.) has its own relative rate

<pre> CACCGGGTCCCCGAGAGCGGGCGCGTGC GCGATCTCACGGACTGACACGTTGACGAGGTTACAGTTGACGTA AAGGAGTGTAGAATGA .....C..... .....TG.....C.....C..... .....G.....C.....AC.....C.....G..... .....C.....C.....C..... T.....C.....C.....C..... .....G.....C.....C.....C..... ..T.....C.....C.....C..... .....C.....C..... .....C.....C.....C..... .....G.....C.....C.....C..... .....G.....C.....C.....C..... .....C.....C.....CG..... </pre>	<pre> ATCTATAAAGTAATAATTTTAGTTGTACATTGCACAAACCTTA .AT..A..GTG..A..AA..T.G.A..TT...A.T..TTTCCG .AT....TT.TT.T.AAA.T.A.A..TT.A.T.T..TTTCCG G.GA.A...AA.T.T.....A...TTT.CTTTT.T..T..C .GAA....AG...T..AC.G.CG..CGTTA.CTT..T..TCC. .AGG....AC...T..A.....C.TTCCT.T..T...C.. .CAAG.G.TA...G...A.G.C.A.G.TTC.TTTTGT..... ..AA.CG.GAC...T..C.....C.TTC.CTC..TG.TA.. ..AG..G.GA...C..C...C...C.TTC.TTT.G...TCCG .AGGGCG.GAA...T..CC...C...C.TT..TTT.GG..TCCG .CA.T...G.CG..C.....AAG...TTC.TTT.....CCG .CAA....CA....GC.A...C.G.AG.GCCT.T.GC...CG ..A.....CG..C.....A.A.C.TTCCTTT..G...CCG </pre>
--	---

$r_1$  applies to subset 1  
 1st+2nd codon positions  
 (sites 1 - 88)

$r_2$  applies to subset 2  
 3rd codon positions  
 (sites 89-132)

Relative rates have mean 1.0:  $r_1 \frac{2}{3} p(r_1) + r_2 \frac{1}{3} p(r_2) = 1$



# Site-specific rates

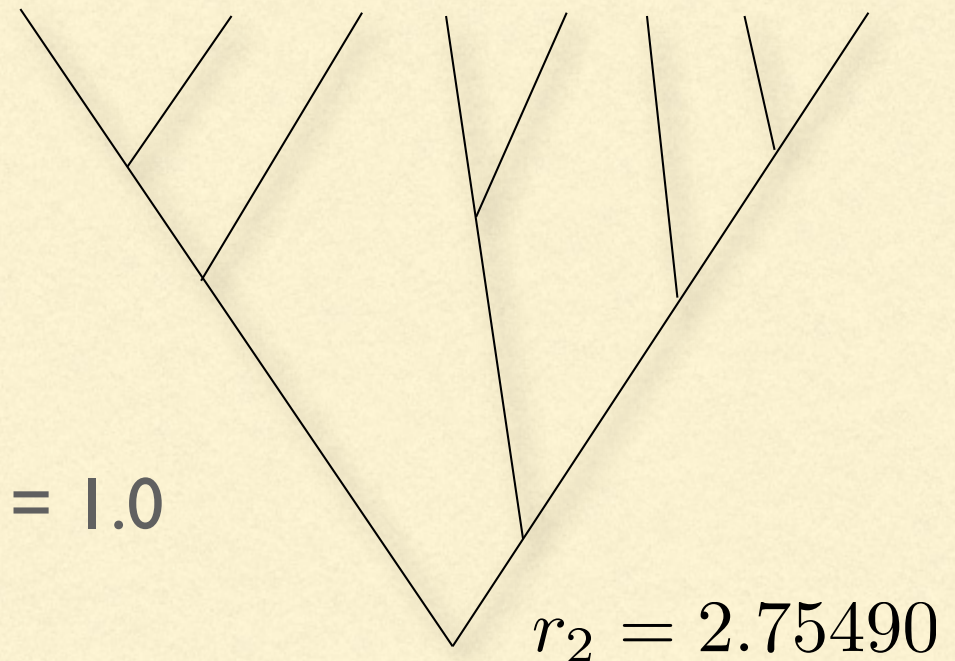
$$L = \underbrace{p(\mathbf{y}_1|r_1) \cdots p(\mathbf{y}_{88}|r_1)}_{\text{1st+2nd codon positions}} \underbrace{p(\mathbf{y}_{89}|r_2) \cdots p(\mathbf{y}_{132}|r_2)}_{\text{3rd codon positions}}$$

▽▽▽

$$r_1 = 0.12255$$

mean relative rate:

$$(0.12255)(2/3) + (2.75490)(1/3) = 1.0$$



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# Site-specific rates

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JC69 transition probabilities that would be used for every site if rate *homogeneity* were assumed:

$$P_{ii}(t) = \frac{1}{4} + \frac{3}{4}e^{-4\beta t} \quad \text{C} \text{ ————— } \text{C}$$

identity

$$P_{ij}(t) = \frac{1}{4} - \frac{1}{4}e^{-4\beta t} \quad \text{C} \text{ ————— } \text{T}$$

difference

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# Site specific rates

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JC69 transition probabilities that would be used for sites in **subset 1**:

$$P_{ii}(t) = \frac{1}{4} + \frac{3}{4}e^{-4r_1\beta t}$$

$$P_{ij}(t) = \frac{1}{4} - \frac{1}{4}e^{-4r_1\beta t}$$

JC69 transition probabilities that would be used for sites in **subset 2**:

$$P_{ii}(t) = \frac{1}{4} + \frac{3}{4}e^{-4r_2\beta t}$$

$$P_{ij}(t) = \frac{1}{4} - \frac{1}{4}e^{-4r_2\beta t}$$

# Mixture models

All  $k$  relative rates applied to every site

```

Q--L--G--V--P--P--E--E--A--G--A--A--V--A--A--E--S--S--T--G--T--W--T--T--V--W--T--D--G--L--T--S--L--D--R--Y--K--G--R--C--Y--H--I--E--
CAACCTGGCGTTCCACCTGAAGAAGCAGGGGCTGCAGTAGCTGCAGAATCTTCTACTGGTACATGGACTACTGTTTGGACTGACGGATTAAGTCTTGGACCGATACAAGGAAGATGCTACGATATTGAA
...A.T.....A.....G.T.G.....A.....A.A.....T.....G.....A.....T.T.....A.....T.....TC.T.T.T.C.C.G
...A.T.....TGT.T.....T.T.....T.....A.A.A.....T.....A.....A.....T.T.....A.C.T.....T.....TC.T.T.T.C.C.G
..G...G.A..G.A.....A.A.....T.....T.....A.....A.....T.TC.T...ACC.T.T.T.T.....TC.....T.G.....C
...G.A.A.....A.G.....T.....A.C.....G.....C.G.....C.T.GC.T.A...C.C.T.T.....TC.....T.C.C...
T...A.G.G.....A.C.....T.....A.....C.....C.T...C.T.C.CC.T...T.....TC.....C.....
...C.A.A.GG..G.....T.A.....G.....A.....G.....C.....A.....G.T...C.T.C...C.T.T.T.T.G.TC.....
...T.A.A.....C.G.....G.A.C.....T.....C.....C.T...C.T.C...C.C.T.C.....TC.G...T.A.....
...A.G.....G.....G.A.....C.....C.....C.....C.T...C.T.C...C.T.T.T.....G.....T.C.C.G
...A.G.G.G.C.G..G.A.A.....T.....C.C.....C.....C.T...C.T...C.T.T.T.....G.GC.....T.C.C.G
...C.A...TG.....G.....C.G.....C.....A.A.G.....T.C.T.C...C.T.T.T.....C.....C.C.C.G
...C.A.A.G.....C.A.....G.C.....A.....C.....G.....A.....G.G.C.CC.T...T.....G.CC.....C.G
...A.....G.....C.G.....C.....C.....A.....A.....C.T...C.T.C.CC.T.T.T.....GC.....CGC.C.G
    
```

site  $i$

$$L_i = p(\mathbf{y}_i|r_1)p(r_1) + p(\mathbf{y}_i|r_2)p(r_2) + \dots + p(\mathbf{y}_i|r_k)p(r_k)$$

Common examples { Invariable sites (I) model  
Discrete Gamma (G) model

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# Invariable sites model (Reeves 1992)

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$$L_i = p(\mathbf{y}_i|r_1)p(r_1) + p(\mathbf{y}_i|r_2)p(r_2)$$

$$L_i = p(\mathbf{y}_i|0.0)p_{\text{invar}} + p(\mathbf{y}_i|r_2)(1 - p_{\text{invar}})$$

# Discrete Gamma model (Yang 1994)

No relative rate is exactly 0.0, and all are equally probable

```

Q--L-G-V-P-P-E-E-A-G-A-A-V-A-A-E-S-S-T-G-T-W--T-T-V-W-T-D-G-L-T-S-L-D-R-Y-K-G-R-C-Y-H-I-E-
CAACCTGGCGTTCCACCTGAAGAAGCAGGGGCTGCAGTAGCTGCAGAATCTTCTACTGGTACATGGACTACTGTTGGACTGACGGATTAAC TAGTTGGACCGATACAAAGGAGATGCTACGATATTGAA
...A.T.....A.....G.T.G.....A.A.....T.....G.....A.....T.T.....A.....T.....TC.T.T.T.C.C.G
...A.T.....TGT.T.T.T.T.....A.A.A.....T.....A.....A.....T.T.....A.....C.T.....T.....TC.T.T.T.C.C.G
...G.....G.A..G.A.....A.A.....T.....T.....A.....A.....T.T.C.T.....ACC.T.T.T.T.....TC.....T.G.....C
...G.A.A.....A.G.....T.....A.C.....G.....C.....G.....C.T.GC.T.A.....C.C.T.T.....TC.....T.C.C...
T..A.G.G.....A.C.....T.....A.....A.....C.T.C.T.C.CC.T.....T.....TC.....C.....
...C.A.A.GG...G...T.A.....G.....A.....G.....C.....A.....G.T...C.T.C...C.T.T.T.G.TC
...T.A.A...C.G...G.A.C.....T.....C.....C.....C.T.C.T.C...C.C.T.C.....TC.G...T.A.....
...A.G...G...G.A.....C.....C.....C.....C.T.C.T.C...C.T.T.T.G.....T.C.C.G
...A.G.G.G.C.G...G.A.A.....T.....C.C.....C.....C.T.C.T.C...C.T.T.T.G.GC...T.C.C.G
...C.A...TG...G...C.G...C.....A.A.G...T.C.T.C...C.T.T.T...C.....C.C.C.G
...C.A.A.G...C.A...G.C...A.....C.....G.A...A...G.G.C.CC.T...T...G.CC...C.G
...A.....C.G...C.....A.....A.....C.T.C.T.C.CC.T.T.T...GC...CGC.C.G
  
```

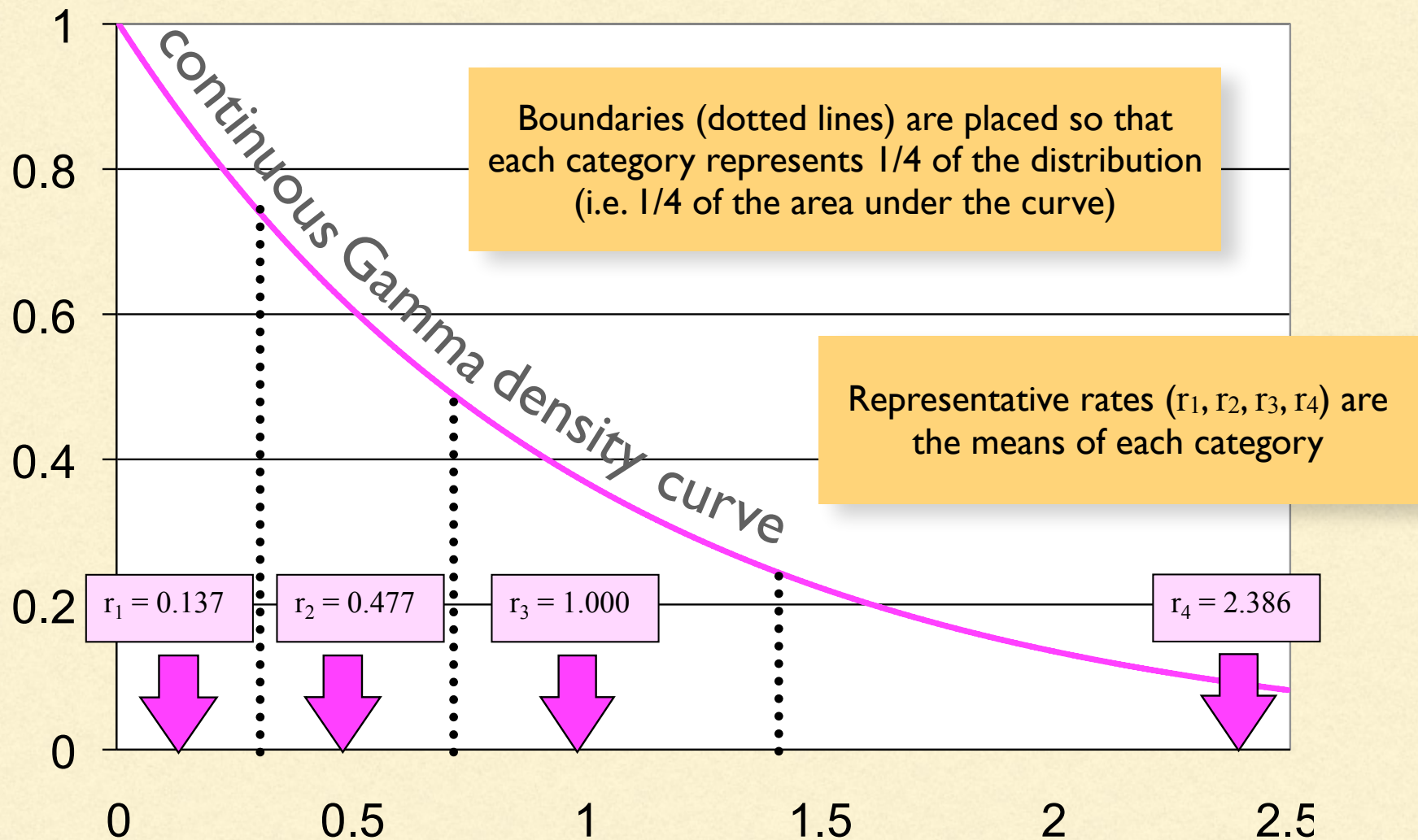
site  $i$

$$L_i = p(\mathbf{y}_i|r_1) \left(\frac{1}{4}\right) + p(\mathbf{y}_i|r_2) \left(\frac{1}{4}\right) + p(\mathbf{y}_i|r_3) \left(\frac{1}{4}\right) + p(\mathbf{y}_i|r_4) \left(\frac{1}{4}\right)$$

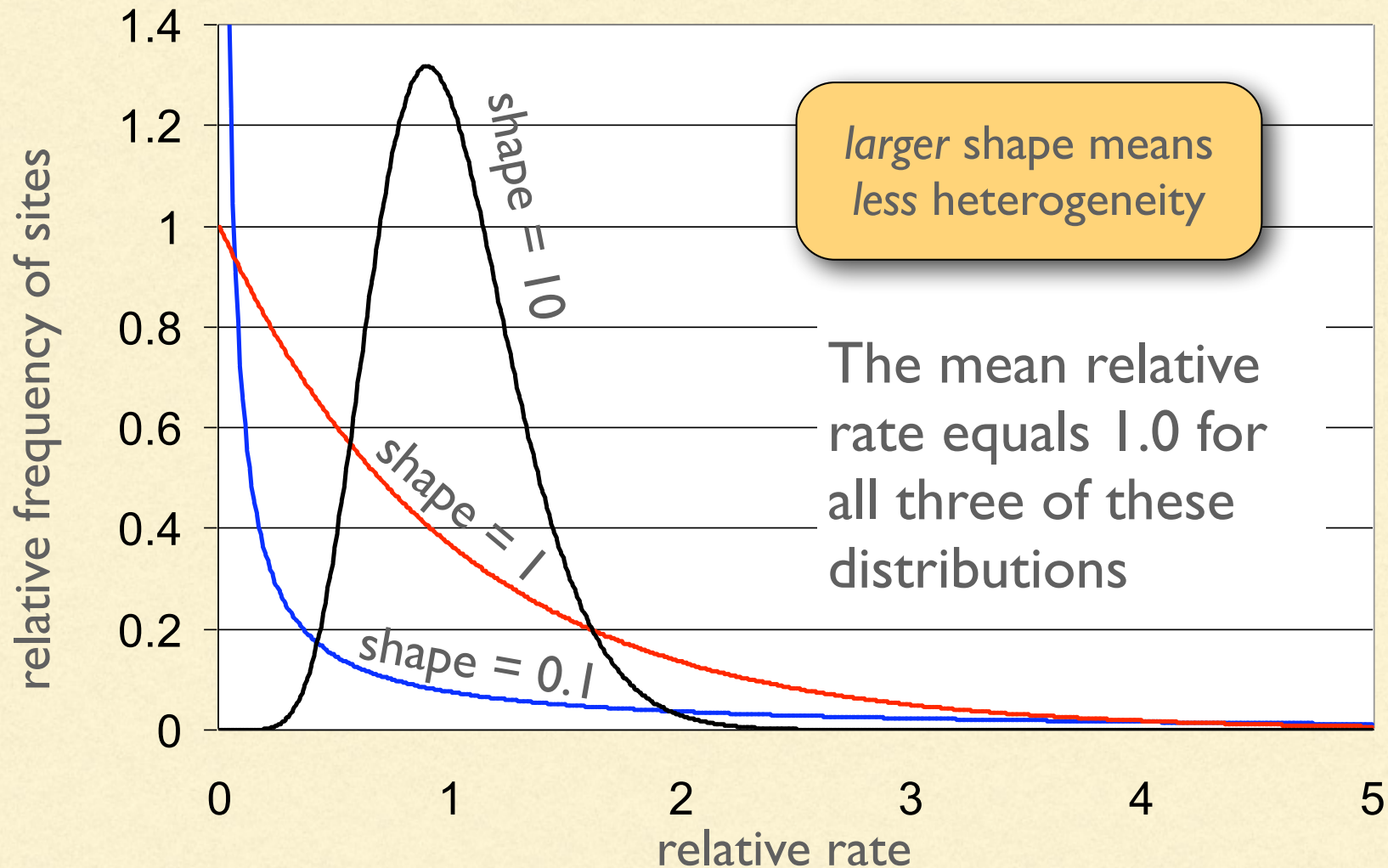
Relative rates are determined by a discrete gamma distribution

Number of rate categories can vary (4 used here)

# Relative rates in 4-category case



# Gamma distributions







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~ Coffee Break ~

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