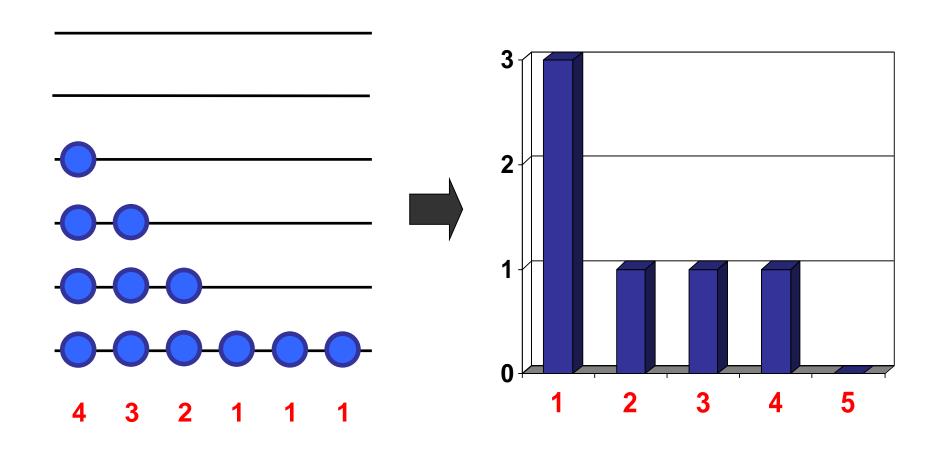
Patterns of Evolution Summary statistics based on segregating sites

Site Frequency Spectrum



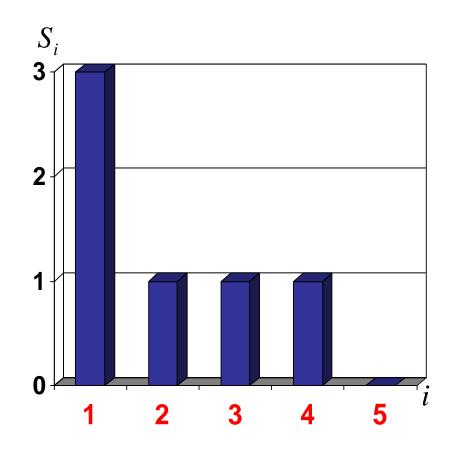
Patterns of Evolution Summary statistics based on segregating sites

Site Frequency Spectrum

$$S_i$$
: number of mutants that appear in i copies in the sample

$$S = \sum_{i=1}^{n-1} S_i : \begin{array}{c} \text{total number of} \\ \text{segregating sites} \\ \text{in an sample of size } n \end{array}$$

$$\pi = \frac{1}{\binom{n}{2}} \sum_{i=1}^{n-1} i(n-i)S_i:$$
 average number of pairwise differences



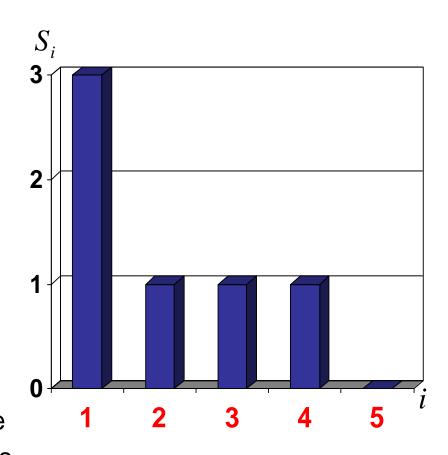
Patterns of Evolution Summary statistics based on segregating sites

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Each mutation of size i contributes to divergence in i $(n-i)$ sequence pairs



Coalescent Theory Estimators

Unbiased estimators of the mutation parameter $\theta = 4Nu$:

Watterson's estimator:

$$\hat{\theta}_W = \frac{S}{a_n} = \sum_{i=1}^{n-1} S_i / \sum_{i=1}^{n-1} \frac{1}{i}$$
 (equal weights)

 π -based estimator:

$$\hat{\theta}_{\pi} = \pi = \binom{n}{2}^{-1} \sum_{i=1}^{n-1} i(n-i) S_i$$
 (intermediate frequencies)

Fay and Wu's estimator:

$$\hat{\theta}_H = \binom{n}{2}^{-1} \sum_{i=1}^{n-1} i^2 S_i$$
 (high frequencies)

singleton estimator:

$$\hat{\theta}_s = \frac{n-1}{n} \left(S_1 + S_{n-1} \right) \quad \text{(extreme frequencies)}$$

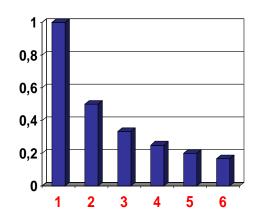
singletons of the folded spectrum

$$D_{T} = \frac{\hat{\theta}_{\pi} - \hat{\theta}_{W}}{\sqrt{\text{Var}[\hat{\theta}_{\pi} - \hat{\theta}_{W}]}}$$

$$D_T = 0$$

$$D_{FL} = \frac{\theta_W - \theta_S}{\sqrt{\text{Var}[\hat{\theta}_W - \hat{\theta}_S]}}$$

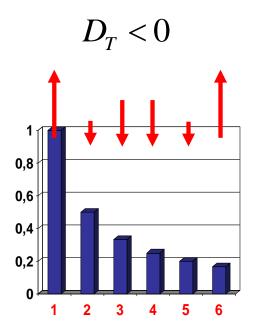
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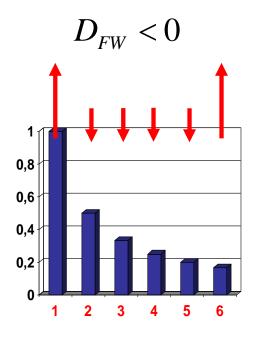
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