

# Joint subtree distributions under two evolutionary tree models

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# OUTLINE

1. INTRODUCTION
2. JOINT SUBTREE DISTRIBUTIONS UNDER TWO EVOLUTIONARY MODELS
3. DISCRIMINATION EFFICIENCY
4. WORK IN PROGRESS

# 1. Introduction

- ▶ Trees provide visual representation of the evolutionary relatedness among organisms, species, or genes.
- ▶ Phylogenetic tree has become a powerful tool in studying evolution and diversification in Biology.
- ▶ It finds wide applications in evolutionary biology, developmental biology (cell lineages) and epidemiology. Such as
  - ▶ Inferring evolutionary process from phylogenetic tree shape<sup>1</sup>;
  - ▶ Studying diversification<sup>2</sup>;
  - ▶ Study of pathogens<sup>3,3a</sup>.
- ▶ Tree shape is the signature of the forces that produce biodiversity<sup>4</sup>
- ▶ Fitting stochastic models to tree data helps infer macro-evolutionary processes such as speciation and extinction rates.

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<sup>1</sup> Mooers & Heard (1997). Inferring evolutionary process from phylogenetic tree shape. *Quart. Review Biol.*, **72**, 31-54.

<sup>2</sup> Morlon (2014). Phylogenetic approaches for studying diversification. *Ecology Letters*, **17**, 508-525.

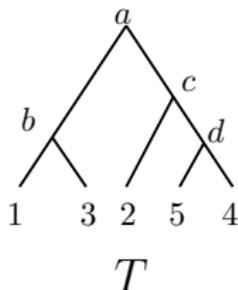
<sup>3</sup> Poon et al. (2013). Mapping the Shapes of Phylogenetic Trees from Human and Zoonotic RNA Viruses. *PLoS*

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<sup>3a</sup> Colijin & Plazzotta (2017). A metric on phylogenetic tree shapes. *Syst. Biol.*, **11**, 113-126.

<sup>4</sup> Mooers & Heard (2002). Using Tree shape. *Sys. Biol.*, **51**, 833-834.

# 1. Introduction



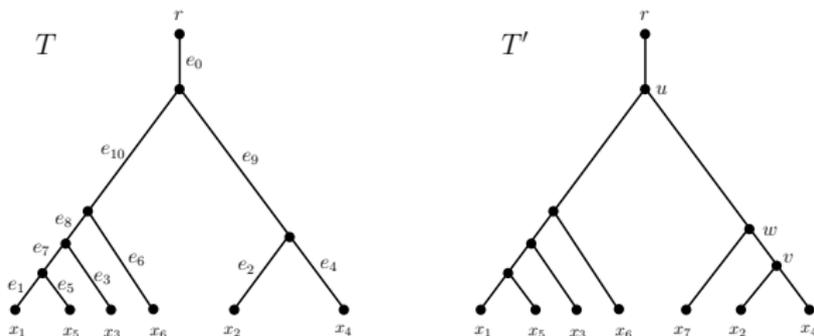
- ▶  $T$  is a phylogenetic tree on  $\mathcal{X} = \{1, \dots, 5\}$ , the set of leaves.
- ▶ All edges are directed away from the root,  $a$  (arrows not drawn).
- ▶ Binary tree  $(V(T), E(T))$ , set of pendant edges denoted by  $E^*(T)$ .
- ▶ Each node is either of  
degree 1 (for leaves);  
degree 2 (root); or  
degree 3 (the rest of internal nodes).

## Our objectives

- ▶ Given a tree  $T$ , to infer if this tree is more likely generated by YHK model than by PDA model.
- ▶ Given a “tree feature” (index), to assess how good this feature in discriminating one generative model from another.
  - ▶ YHK trees are observed to be more symmetric. Many tree indices have been proposed to measure tree balance.
  - ▶ Examples: Colless index and Sackin index

## 1.2 Two random generative tree models: YHK & PDA

- ▶ Two common generative tree models: YHK (Yule-Harding-Kingsman) model and PDA (Proportional-to-Distinguishable-Arrangements) model



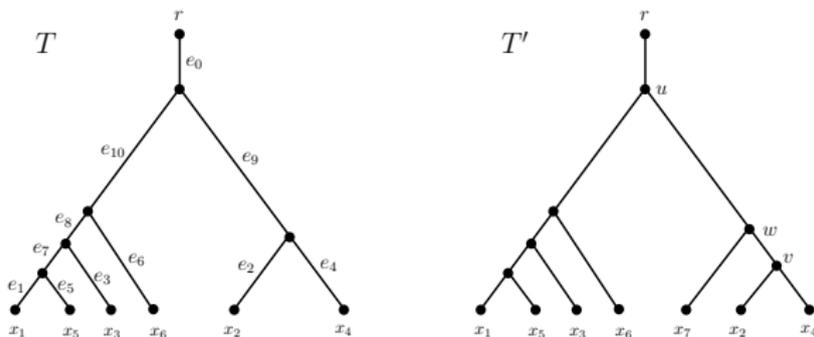
- ▶ YHK: Choose a pendant edge at random in  $T$  to form  $T'$ .
- ▶ PDA: Choose an edge at random from all edges in  $T$  to form  $T'$ .
- ▶ In this figure, the PDA model:  $T' = T[e_9; x_7]$  is obtained from  $T$  by attaching the leaf labelled  $x_7$  to edge  $e_9$ .

## 1.2 Two generative tree models: YHK & PDA

Starting with a tree of 2 leaves, iteratively attach (manner to be specified) one leaf at a time to grow the tree to its desired size.

- ▶ YHK, also known as equal-rates-Markov model: all extant lineages have equal probabilities of speciating at an instant.
  - ▶ Uniformly sample a pendant edge from the tree's pendant edges, and attach a new leaf to it.
- ▶ PDA: Each “possible arrangement” of  $n$  leaves into a tree is equally likely.
  - ▶ Uniformly sample an edge from the present tree's edges, and attach a new leaf to it.

## 2.1 Subtree



- ▶ A cherry—a subtree with two leaves.

For examples,

$\{x_1, x_5\}$  (together with their parent) forms a cherry, and we think of three edges:  $e_7, e_1, e_5$

$\{x_2, x_4\}$  forms another cherry with edges:  $e_9, e_2, e_4$ .

- ▶ A pitchfork—a subtree with three leaves.

E.g.,  $\{x_1, x_5, x_3\}$  with their most recent common ancestor; 5-edge:  $e_8, e_7, e_3, e_1, e_5$ .

## 2.2 Subtrees of sizes 2 and 3 (Cherry and Pitchfork)

- ▶ It has been observed that YHK generated trees are more “symmetric”: and hence contain more cherries.

### Definitions

- ▶  $A_n$ : Number of pitchforks in a tree with  $n$  leaves
- ▶  $C_n$ : Number of cherries in a tree with  $n$  leaves
- ▶ Let  $h_n$  denote the joint pmf of  $A_n$  and  $C_n$  under the YHK model. That is,

$$h_n(a, c) = P_Y[A_n = a, C_n = c].$$

- ▶ Let  $f_n$  be the pmf of  $A_n$ ; and  $g_n$  pmf of  $C_n$  under the YHK model.
- ▶ Denote  $\tilde{h}_n, \tilde{f}_n, \tilde{g}_n$  the counterparts under the PDA model.

**Thm 1** We have the following recurrence relations.

(a) YHK Model

For  $n \geq 3$ ,

$$h_{n+1}(a, c) = \frac{2a}{n} h_n(a, c) + \frac{a+1}{n} h_n(a+1, c-1) \\ + \frac{2(c-a+1)}{n} h_n(a-1, c) + \frac{n-a-2c+2}{n} h_n(a, c-1);$$

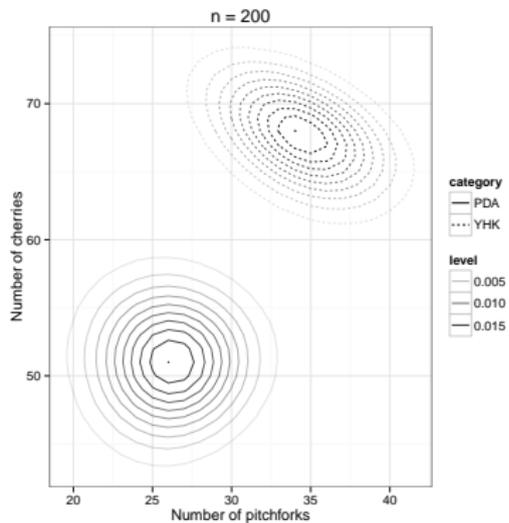
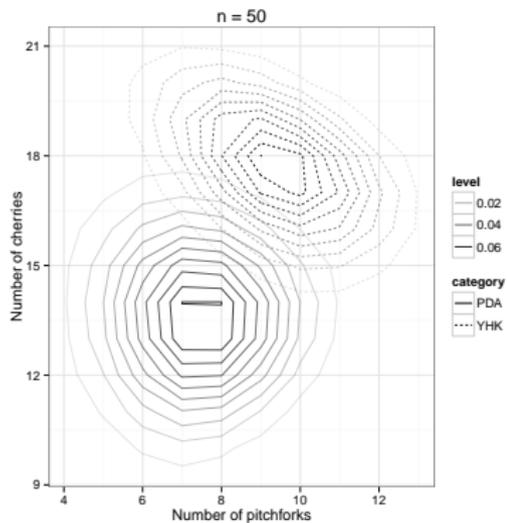
and  $h_3(1, 1) = 1$ , and  $h_3(a, c) = 0$  if  $(a, c) \neq (1, 1)$ .

(b) PDA Model

For  $n \geq 3$ ,

$$\tilde{h}_{n+1}(a, c) = \frac{n+3a-c-1}{2n-1} \tilde{h}_n(a, c) + \frac{a+1}{2n-1} \tilde{h}_n(a+1, c-1) \\ + \frac{3(c-a+1)}{2n-1} \tilde{h}_n(a-1, c) + \frac{n-a-2c+2}{2n-1} \tilde{h}_n(a, c-1);$$

and  $\tilde{h}_3(1, 1) = 1$ , and  $\tilde{h}_3(a, c) = 0$  if  $(a, c) \neq (1, 1)$ .



## Key step in the proof of Thm 1

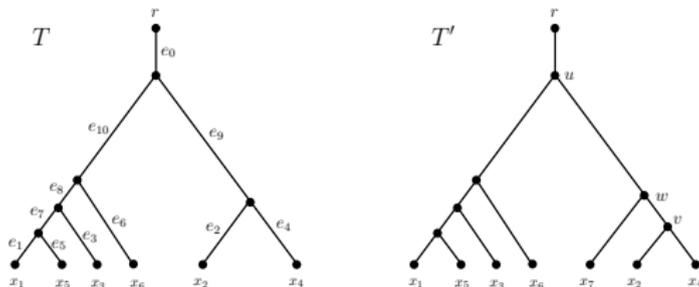
Edge decomposition for a PDA tree: Partition  $E(T)$  into

$E_1(T)$ : pendant edges that are contained in a pitchfork but not in a cherry;

$E_2(T)$ : edges that are contained in a cherry but not in a pitchfork;

$E_3(T)$ : pendant edges that are contained in neither a pitchfork nor a cherry;

$E_4(T) = E(T) \setminus (E_1(T) \cup E_2(T) \cup E_3(T))$ .



$$E_1(T) = \{e_3\}, \quad E_2(T) = \{e_2, e_4, e_9\}, \quad E_3(T) = \{e_6\}, \\ E_4(T) = \{e_0, e_1, e_5, e_7, e_8, e_{10}\}.$$

## Key step in the proof of Thm 1

Edge decomposition for a PDA tree

- ▶ Track the changes in the numbers of pitchforks and cherries when a new leaf join the tree  $T$  at  $E_i(T)$ :

$$(A(T'), C(T')) = \begin{cases} (A(T) - 1, C(T) + 1), & e \in E_1(T), \\ (A(T) + 1, C(T)), & e \in E_2(T), \\ (A(T), C(T) + 1), & e \in E_3(T), \\ (A(T), C(T)), & e \in E_4(T), \end{cases}$$

and

$$\begin{aligned} |E_1(T)| &= A(T), \\ |E_2(T)| &= 3(A(T) - C(T)), \\ |E_3(T)| &= n - A(T) - 2C(T), \\ |E_4(T)| &= n - 1 + 3A(T) - C(T). \end{aligned}$$

**Thm 2** Denote expectation under YHK (resp., PDA) model by  $\mathbb{E}_Y$  (resp.,  $\mathbb{E}_U$ ). Let  $\phi : \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ . Then, for  $n \geq 4$ ,

$$\begin{aligned} n\mathbb{E}_Y[\phi(A_{n+1}, C_{n+1})] &= 2\mathbb{E}_Y[A_n\phi(A_n, C_n)] \\ &\quad + \mathbb{E}_Y[A_n\phi(A_n - 1, C_n + 1)] \\ &\quad + 2\mathbb{E}_Y[(C_n - A_n)\phi(A_n + 1, C_n)] \\ &\quad + \mathbb{E}_Y[(n - A_n - 2C_n)\phi(A_n, C_n + 1)]; \end{aligned}$$

$$\begin{aligned} (2n - 1)\mathbb{E}_U[\phi(A_{n+1}, C_{n+1})] &= \mathbb{E}_U[(n + 3A_n - C_n - 1)\phi(A_n, C_n)] \\ &\quad + \mathbb{E}_U[A_n\phi(A_n - 1, C_n + 1)] \\ &\quad + 3\mathbb{E}_U[(C_n - A_n)\phi(A_n + 1, C_n)] \\ &\quad + \mathbb{E}_U[(n - A_n - 2C_n)\phi(A_n, C_n + 1)]. \end{aligned}$$

## Special cases

(1) Let  $\psi : \mathbb{R} \rightarrow \mathbb{R}$  be any function.

(a) Define  $\phi(x, y) = \psi(y)$ , we have

$$\begin{aligned}n\mathbb{E}_Y[\psi(C_{n+1})] &= \mathbb{E}_Y[2C_n\psi(C_n) + (n-2C_n)\psi(C_n+1)]; \\(2n-1)\mathbb{E}_U[\psi(C_{n+1})] &= \mathbb{E}_U[(n+2C_n-1)\psi(C_n) \\ &\quad + (n-2C_n)\psi(C_n+1)].\end{aligned}$$

(b) Define  $\phi(x, y) = I_k(y)$ , we have

$$\begin{aligned}ng_{n+1}(k) &= 2k g_n(k) + (n-2k+2)g_n(k-1); \\(2n-1)\tilde{g}_{n+1}(k) &= (n+2k-1)\tilde{g}_n(k) + (n-2k+2)\tilde{g}_n(k-1),\end{aligned}$$

for  $n \geq 3$  and  $k \geq 1$ .

(c) Take  $\phi(x, y) = y$  to derive  $\mathbb{E}_Y[C_n]$  and  $\mathbb{E}_U[C_n]$ ;

(d)  $\phi(x, y) = y^2$  to derive  $\text{Var}_Y[C_n]$  and  $\text{Var}_U[C_n]$ .

## On the average

- ▶ more cherries with less variation in a YHK tree than in a PDA tree:

$$\begin{aligned}\mathbb{E}_Y[C_n] &= \frac{n}{3} > \mathbb{E}_U[C_n] = \frac{n(n-1)}{2(2n-3)} \sim \frac{n}{4}; \\ \text{Var}_Y[C_n] &= \frac{2n}{45} < \text{Var}_U[C_n] = \frac{n(n-1)(n-2)(n-3)}{2(2n-3)^2(2n-5)} \sim \frac{n}{16};\end{aligned}$$

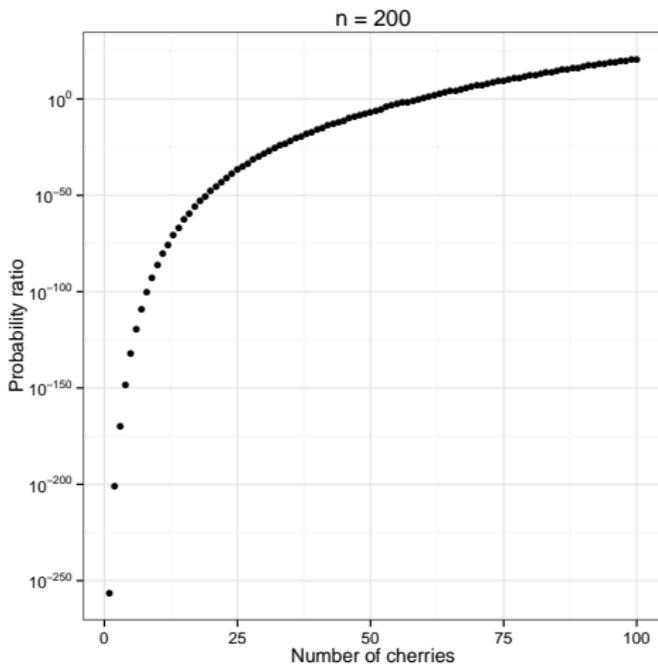
- ▶ more pitchforks but also with more variation in a YHK tree than in a PDA tree:

$$\begin{aligned}\mathbb{E}_Y[A_n] &= \frac{n}{6} > \mathbb{E}_U[A_n] = \frac{n(n-1)(n-2)}{2(2n-3)(2n-5)} \sim \frac{n}{8}; \\ \text{Var}_Y[A_n] &= \frac{23n}{420} > \text{Var}_U[A_n] = \frac{3n(n-1)(n-2)(n-3)(4n^3-40n^2+123n-110)}{4(2n-3)^2(2n-5)^2(2n-7)(2n-9)} \sim \frac{3n}{64};\end{aligned}$$

- ▶ Correlation of numbers of cherries and pitchforks is constant (independent of  $n$ ) under YHK; whereas correlation =  $O(1/n)$  under PDA:

$$\text{Cor}_Y(A_n, C_n) = -\sqrt{\frac{14}{69}}; \quad \text{Cor}_U(A_n, C_n) \sim \frac{-1}{\sqrt{3}n}.$$

We examine the likelihood ratio of the number of cherries under YHK versus PDA:  $f_n(k)/\tilde{f}_n(k)$ . It is monotone in  $k$  for fixed  $n$ .



### 3.1 Model-based total variation distance between YHK and PDA

- ▶ **Definition** We define a total variation distance between YHK model and PDA model on the collection of trees with  $n$  leaves as

$$mtv_n(YHK, PDA) := \frac{1}{2} \sum_{T \in \mathcal{T}_n} |P_Y(T) - P_U(T)|.$$

- ▶ For  $T \in \mathcal{T}_n$ , it is known

$$P_U(T) = \frac{1}{(2n-3)!!} \quad \text{and} \quad P_Y(T) = \frac{2^{n-1}}{n! \pi(T)}$$

where

$$\pi(T) := \prod_{v \in \overset{\circ}{V}} (\Lambda(v) - 1)$$

where  $\overset{\circ}{V}$  denotes the set of internal nodes of  $T$ , and  $\Lambda(v)$  is the number of descendants of  $v$  (Semple & Steel, 2003).

- ▶ An application:  $H_0 : YHK$  vs  $H_1 : PDA$ ,

$$\text{power} \leq \alpha + mtv_n(YHK, PDA).$$

- ▶ By exhaustive enumeration, we computed the numerical values of  $mtv_n(YHK, PDA)$  for  $n = 3, \dots, 20$ .
- ▶ For larger  $n$ , we estimate  $mtv_n(YHK, PDA)$  by simulation.

Rewrite

$$\begin{aligned}
 mtv_n(YHK, PDA) &= \sum_{T \in \mathcal{T}_n} [P_Y(Y) - P_U(T)]_+ \\
 &= \sum_{T \in \mathcal{T}_n} \left[ \frac{P_Y(Y)}{P_U(T)} - 1 \right]_+ P_U(T) = \mathbb{E}_U[R]
 \end{aligned}$$

where  $R : \mathcal{T}_n \rightarrow \mathbb{R}$  defined as  $R(T) = \left[ \frac{P_Y(Y)}{P_U(T)} - 1 \right]_+$ .

Sample  $T_1, T_2, \dots, T_m$  from  $\mathcal{T}_n$  according to the PDA model. Compute  $R(T_1), R(T_2), \dots, R(T_m)$ . And

$$\frac{1}{m} \sum_{i=1}^m R(T_i) \approx \mathbb{E}_U[R] = mtv_n(YHK, PDA).$$

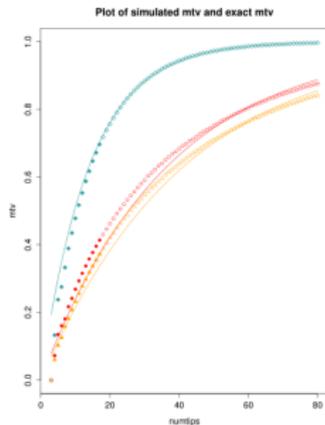


FIGURE 1. Exact and approximate total variation distances between tree models. Blue diamonds, red circles and orange triangles denote  $mtv(n; YHK, PDA)$ ,  $mtv(n; YHK, AB)$  and  $mtv(n; PDA, AB)$  respectively. Solid plotting characters and hollow plotting characters indicate exact values and approximate values by simulation respectively. The lines denote the fitted curves by regression.

- ▶ Simulation is conducted up to  $n = 80$ .
- ▶ We plot  $-\log(1 - mtv_n(YHK, PDA))$  as a function of  $n$ , suggesting

$$mtv_n(YHK, PDA) \approx 1 - e^{-\alpha - \beta n}.$$

## 3.2 Discrimination Efficiency

- ▶ For a chosen feature of a tree  $T \in \mathcal{T}_n$ , we propose a score, discrimination efficiency (DE), to assess the extent this feature is able to discriminate *YHK* from *PDA*.
- ▶ For examples,
  - ▶ Colless index of balanced tree

$$\text{Colless}(T) = \sum_{v \in \overset{\circ}{V}} |R(v) - L(v)|$$

where  $\overset{\circ}{V}$  denotes the set of interior vertices;  $R(v)$  (resp.,  $L(v)$ ) denotes the number of descendants by the right (left) child of  $v$ .

- ▶ Number of cherries in  $T$ ,  $C_n(T)$
- ▶ Joint numbers of cherries and pitchforks in  $T$ ,  $(A_n(T), C_n(T))$

## 3.2 Discrimination Efficiency

- ▶ Suppose we consider a tree index  $I_n$  on  $\mathcal{T}_n$

$$\begin{aligned} \text{mtv}_{I_n}(YHK, PDA) &:= d_{tv}(\mathcal{L}(I_n|YHK), \mathcal{L}(I_n|PDA)) \\ &= \frac{1}{2} \sum_a |g_n(a) - \tilde{g}_n(a)| \end{aligned}$$

where  $g_n(a) = \sum P_Y(T)$  where the sum is taken over all  $T \in \mathcal{T}_n$  with  $I_n(T) = a$ .

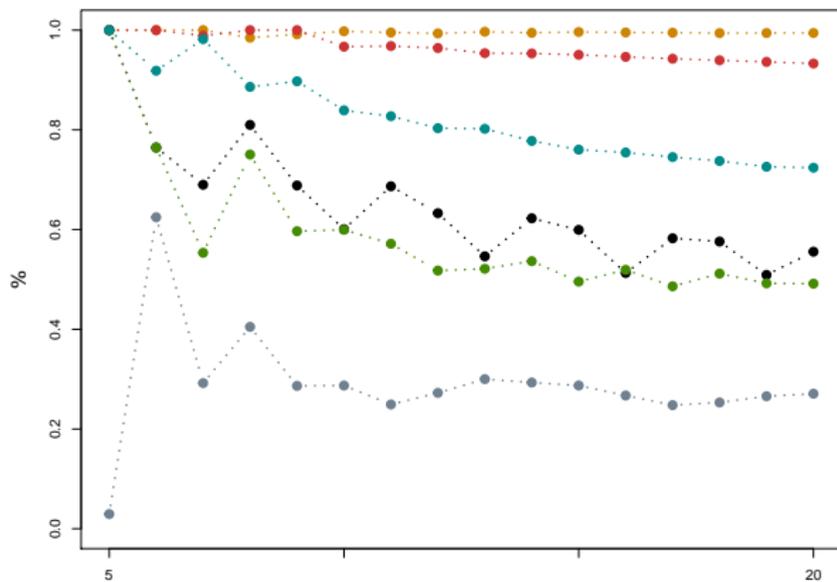
- ▶ Triangle inequality implies

$$0 \leq \text{mtv}_{I_n}(YHK, PDA) \leq \text{mtv}_n(YHK, PDA).$$

- ▶ Define discrimination efficiency of index  $I_n$ :

$$DE(C_n) := \text{mtv}_{C_n}(YHK, PDA) / \text{mtv}_n(YHK, PDA) \in [0, 1].$$

### Index Total Variation



- prod(lambda)
- cherries
- pitchforks
- joint\_cherry\_pitchfork
- symnodes
- colless

## 3.2 Discrimination Efficiency

- ▶ Write  $\mathcal{T}_n(a) = \{T \in \mathcal{T}_n : I_n(T) = a\}$ .

- ▶ Since

$$mtv_{I_n}(YHK, PDA) = \frac{1}{2} \sum_a \left| \sum_{T \in \mathcal{T}_n(a)} [P_Y(T) - P_U(T)] \right|,$$

index  $I_n$  is optimal

$\Leftrightarrow$  for each  $a$ ,  $P_Y(T) - P_U(T)$  is of the same sign for all  $T \in \mathcal{T}_n(a)$

$\Leftrightarrow$  for each  $a$ ,  $P_Y(T)$  is a constant

$\Leftrightarrow$  for each  $a$ ,  $\pi(T) = \prod_{v \in V} (\Lambda(v) - 1)$  is a constant

- ▶ This leads to the tree index:  $\sum_{v \in V} \log(\Lambda(v) - 1)$  one that is proposed by Blum and Francois (2006)<sup>5</sup>.

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<sup>5</sup>Which random processes describe the tree of life? A large scale study of phylogenetic tree imbalance. *Sys. Biol.*, **55**, 685-691

## 4 Work in progress

### 4.1 Extend the results to unrooted trees

- ▶ Studying unrooted trees is of interest: many inference methods return unrooted trees first. Then some ways to root this tree if a rooted tree is desired.
- ▶ C + Thompson + Wu observed the edge decomposition step can be modified to handle unrooted tree.

## 4 Work in progress

### 4.2 We are interested in the correlation of the number of nodes with $a$ -descendants and the number of nodes of $b$ -descendants

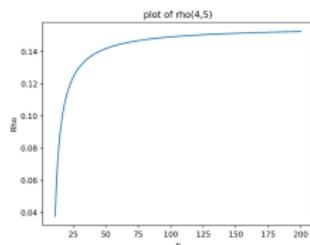
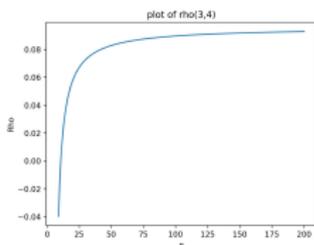
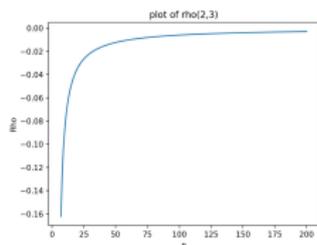
Number of cherries = Number of nodes with 2-descendants;

Number of pitchforks = Number of nodes with 3-descendants.

- ▶ Recall YHK:  $\rho_{n, YHK}(2, 3)$  is a constant in  $n$ .

C + Wu proved that the same is true for general  $a$  and  $b$ .

Under PDA model: correlation can be positive!



Left panel:  $a = 2, b = 3$ , the correlation is negative for all  $n$ .

Middle panel:  $a = 3, b = 4$ , the correlation changes sign from -ve to +ve.

Right panel:  $a = 4, b = 5$ , the correlation is positive for all  $n$ .

C + Wu gave a characterization for which pair  $(a, b)$  will lead to left panel and right panel.

## 4 Work in progress

### 4.3 Two one-parameter families of evolutionary models

#### 4.3.a. Aldous' $\beta$ -branch split model (1996, 2001)

- ▶ Prob of left sister clade and right sister clade contain  $i$  leaves and  $n - i$  leaves respectively is

$$p_{\beta}(i|n) = \frac{1}{a_n(\beta)} \frac{\Gamma(\beta + i + 1)\Gamma(\beta + n - i + 1)}{\Gamma(i + 1)\Gamma(n - i + 1)}, \quad 1 \leq i \leq n - 1.$$

- ▶  $\beta = 0$ : YHK
- ▶  $\beta = -\frac{3}{2}$ : PDA
- ▶ Yule model did not fit the empirical tree data well.
- ▶ Many empirical trees are consistent with Aldous' branch split model:  $\beta \approx -1$ .<sup>6</sup>

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<sup>6</sup>AB model in previous plot corresponds to  $\beta = -1$

## 4 Work in progress

### 4.3 Two one-parameter families of generative models

#### 4.3.b. Ford's $\alpha$ model (2005)

- ▶ Starting with a small tree, grow it to the desired size by adding one leaf at a time.
- ▶ Fix  $\alpha \in [0, 1]$ .  
Attach a new leaf to the current tree's internal edge with probability  $\alpha$ ; and to a pendant edge  $1 - \alpha$ .
- ▶ YHK:  $\alpha = 0$ ;  
PDA:  $\alpha = \frac{1}{2}$ .