

# What does the tree of life look like as it grows?

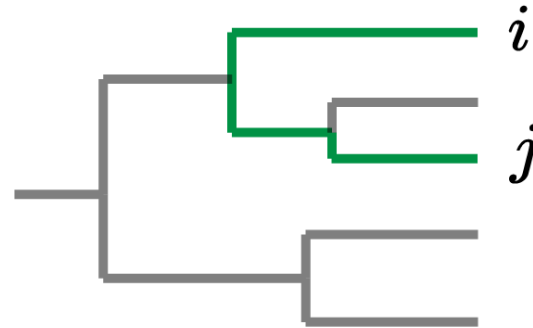
**Toward a dynamic theory of phylogenetic space**

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# Dimensions of phylogenetic space

Branch length  $\propto$  time

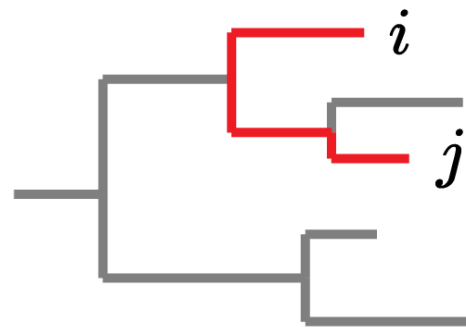


$$d : \mathbb{R}^2 \longrightarrow \mathbb{R}$$

$$d_t(i, j) = \Delta t \text{ [time]}$$

Chronogram

Branch length  $\propto$  form



$$d : \mathbb{R}^2 \longrightarrow \mathbb{R}$$

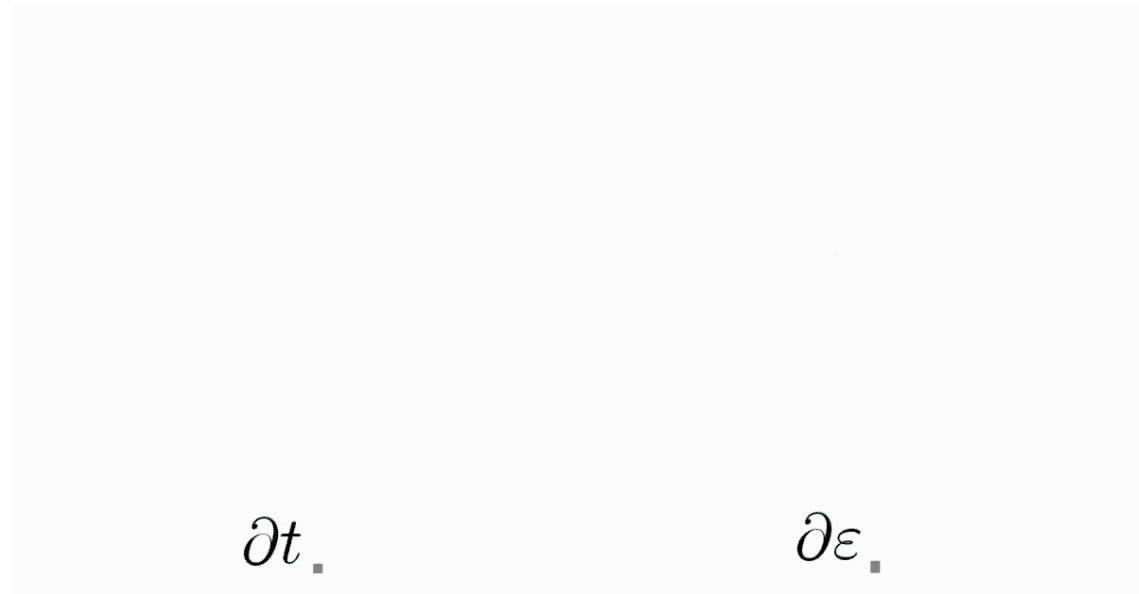
$$d_\varepsilon(i, j) = \Delta \varepsilon \text{ [form]}$$

Phylogram

These are **static trees**. We have the quantities  $\Delta t$  and  $\Delta \varepsilon$ , but not  $\partial t$  and  $\partial \varepsilon$ .

Chronograms and phylograms are **individually inadequate** to model the dynamic case

# Requirement of sufficient dimensions:



$\partial t$

$\partial \varepsilon$

Chronogram: no  $\partial \varepsilon$

Phylogram: no  $\partial t$

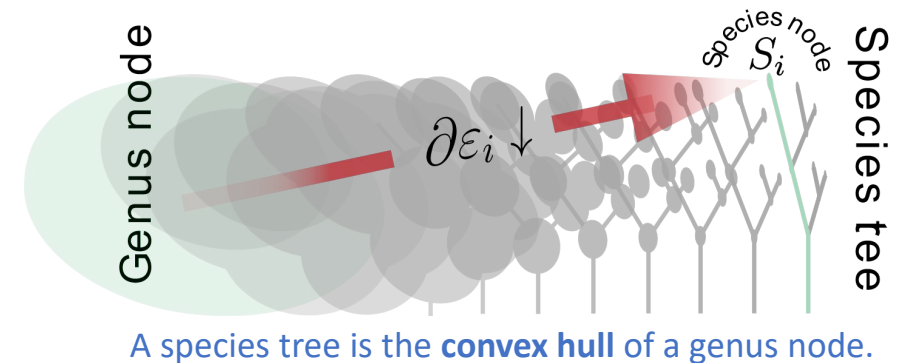
Each has **insufficient dimensions** to represent the dynamic  $\frac{\partial \varepsilon}{\partial t}$ .

Instead, need **two metrics**  $d_t(i, j)$  and  $d_\varepsilon(i, j)$  defined on the same space.

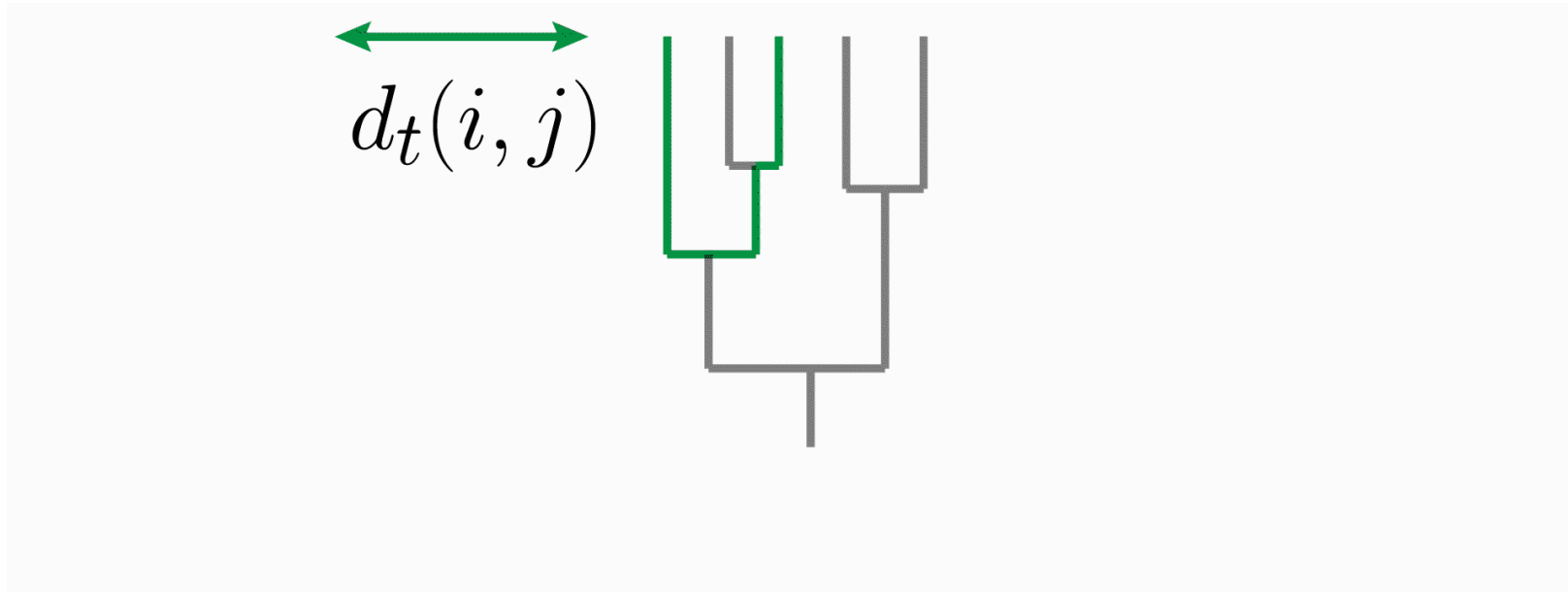
That is, to represent the dynamic  $\frac{\partial \varepsilon}{\partial t}$ , we must **combine chronograms and phylograms**.

# Requirements of a dynamic model

- Sufficient dimensions for the dynamic  $\frac{\partial \varepsilon}{\partial t}$
- The random principle
  - The foundation of modern biology
- The nested principle
  - The logical structure of descent-with-modification
- The organism-population duality principle
  - Mantra in biology: “populations, not organisms, evolve”.



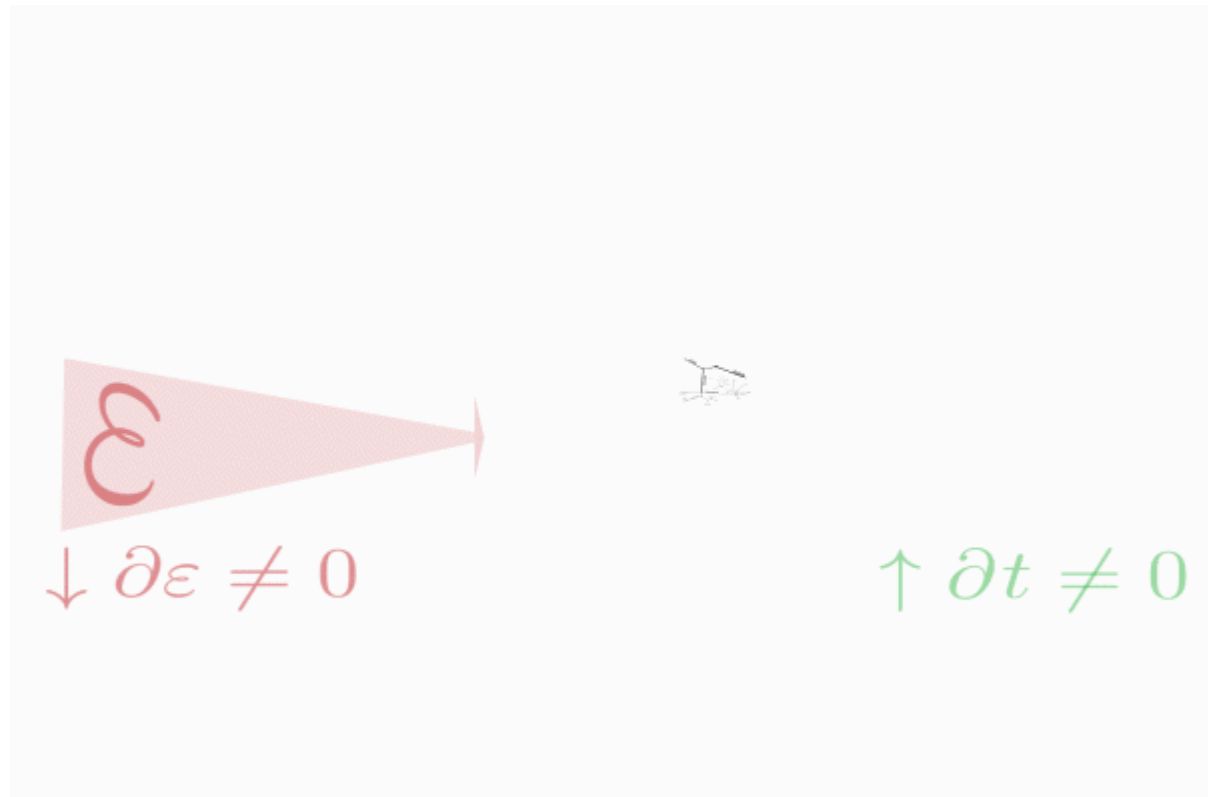
Hint: a tree is not needed if only one metric



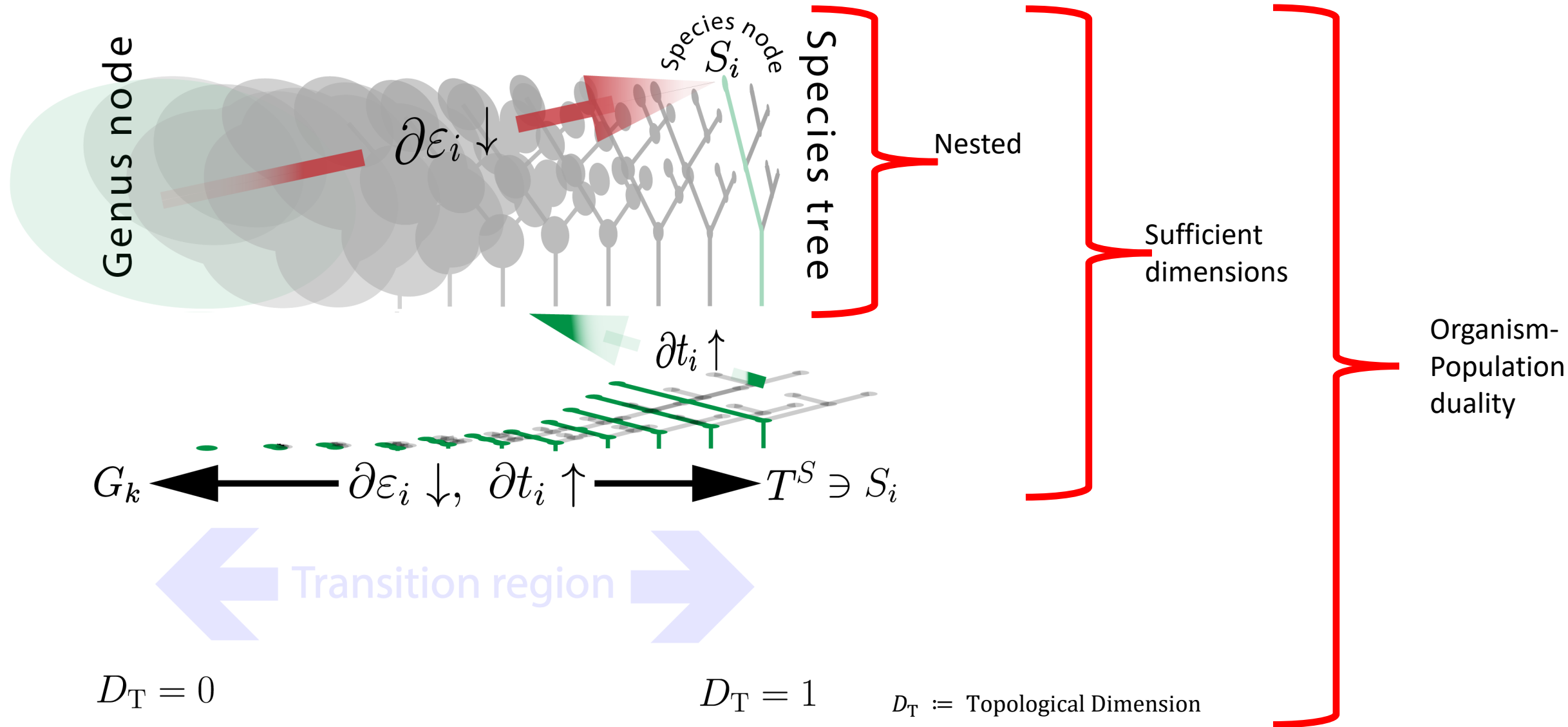
A tree in  $\mathbb{R}^2$  with only one metric can be reduced to  $\mathbb{R}$  if we utilize **scale**.  
(Think random Cantor dust)

# Approach: assign one variable to scale, the other to branch length

Form  $\varepsilon \rightarrow$  scale  
Time  $t \rightarrow$  branch length  
Evolution  $\partial\varepsilon \rightarrow$  scale contraction  
Time expansion  $\partial t \rightarrow$  branch elongation



# The dynamic: $\partial \varepsilon_i / \partial t_i$



# The model: trees within trees within trees...



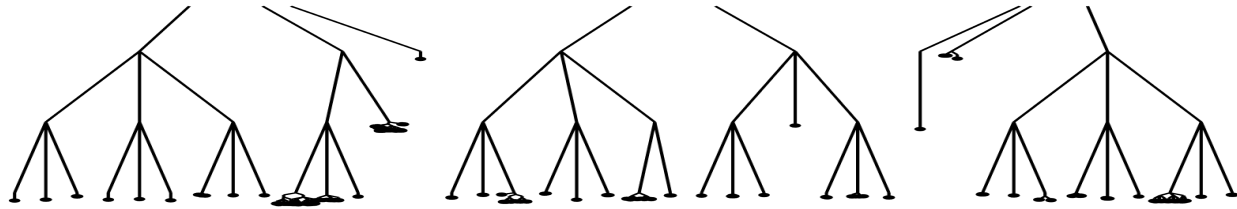
$\partial \varepsilon_{\dots i}$  : path-specific contraction on the unit square to extant/living leaf

$\partial t_{\dots i}$  : path-specific branch elongation

$$\frac{\partial \varepsilon_{\dots i}}{\partial t_{\dots i}}$$



# Stochastic Iterated Function System



## Auxiliary functions

Let  $R: [0,1] \times [0,1] \rightarrow [0,1] \times [0,1]$  such that

$$R(\mathcal{T}_{0\dots i}) = (a, b), \text{ the coordinates of the root node of tree } \mathcal{T}_{0\dots i}.$$

Let  $L: [0,1] \rightarrow \mathbb{Z}$  such that

$$L(\mathcal{T}_{0\dots i}) = \text{the number of leaves (terminal nodes) of tree } \mathcal{T}_{0\dots i}, \text{ and}$$

Let  $l_{ij}$  denote the  $j$ th leaf of tree  $\mathcal{T}_{0\dots i}$ .

## Ratio list:

For tree  $\mathcal{T}_{0\dots i}$  in the  $K - 1^{\text{th}}$  system iterate:

$$\left( \varepsilon_{0\dots i_1}, \dots, \varepsilon_{0\dots i_{L(\mathcal{T}_{0\dots i}^{K-1})}} \right) \quad \text{Assigned to biological form}$$

Where each  $\varepsilon_{0\dots i_j}$  is a random variable with  $\varepsilon_{0\dots i_j} \sim U(0, \varepsilon_{0\dots i})$ , and  $\varepsilon_1 \sim U(0,1)$ .

## Function list:

The leaves  $l_{ij}$  of tree  $\mathcal{T}_{0\dots i}$  in the  $K^{\text{th}}$  system iterate  $E_K$ , are comprised of

$L(\mathcal{T}_{0\dots i}^{K-1})$  executions of function  $T_{0\dots i_j}^K$ :

$$\left( T_{0\dots i_1}^K, \dots, T_{0\dots i_{L(\mathcal{T}_{0\dots i}^{K-1})}}^K \right)$$

Where  $i \in \mathbb{Z}$  with range  $1 \leq i \leq L(\mathcal{T}^{K-1})$

Here,  $T_{0\dots i_j}^K: [0,1] \times [0,1] \rightarrow [0,1] \times [0,1]$  is given by:

$$T_{0\dots i_j}^K = \mathbf{Z} * \varepsilon_{0\dots i_j} + \|R(T_{0\dots i_j}^K) - l_{ij}\|_2, l_{ij} \in \mathcal{T}_{0\dots i}^{K-1}$$

**GW Scale** **Shift**

Where  $\mathbf{Z}$  is a sequence generated by the Galton-Watson branching process:

$$Z_{p+1} = \begin{cases} \xi_1^{p+1} + \dots + \xi_{Z_p}^{p+1}, & Z_p > 0 \\ 0, & Z_p = 0 \end{cases}$$

Static approximation tree

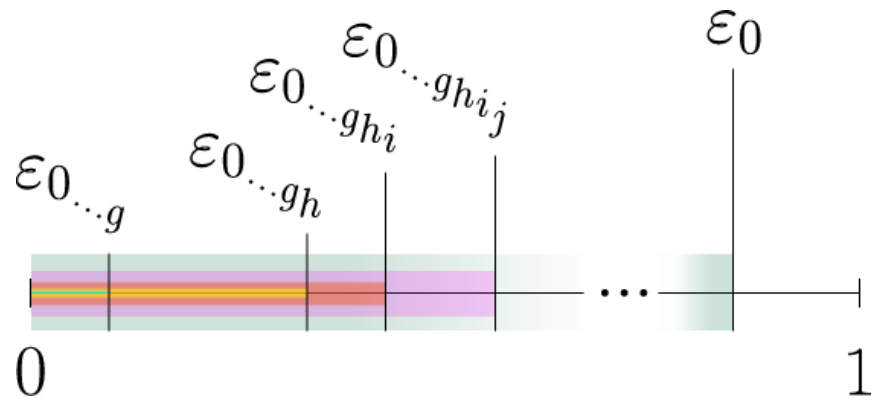
With  $\xi_r^p \in \mathbb{Z}$  i.i.d. Nonnegative random variables, and  $p, r \geq 0$ .

# Every scale path $\varepsilon_{\dots i_j}$ is a Markov chain

Each  $\varepsilon_{\dots i_j}$  is a random variable with  $\varepsilon_{\dots i_j} \sim U(0, \varepsilon_{\dots i})$ , and  $\varepsilon_1 \sim U(0,1)$ . So  $\varepsilon_{\dots i_j} < \varepsilon_{\dots i}$  for all  $\varepsilon_{\#}$

$$PR(\varepsilon_{0\dots h_{ij}} = \varepsilon_{0\dots h_{ij}} \mid \varepsilon_{0\dots h_i} = \varepsilon_{0\dots h_i}, \varepsilon_{0\dots h} = \varepsilon_{0\dots h}, \dots, \varepsilon_0 = \varepsilon_0) = PR(\varepsilon_{0\dots h_{ij}} = \varepsilon_{0\dots h_{ij}} \mid \varepsilon_{0\dots h_i} = \varepsilon_{0\dots h_i})$$

**Non-stationary.**  
But very well-behaved.



Therefore, the entropy is:

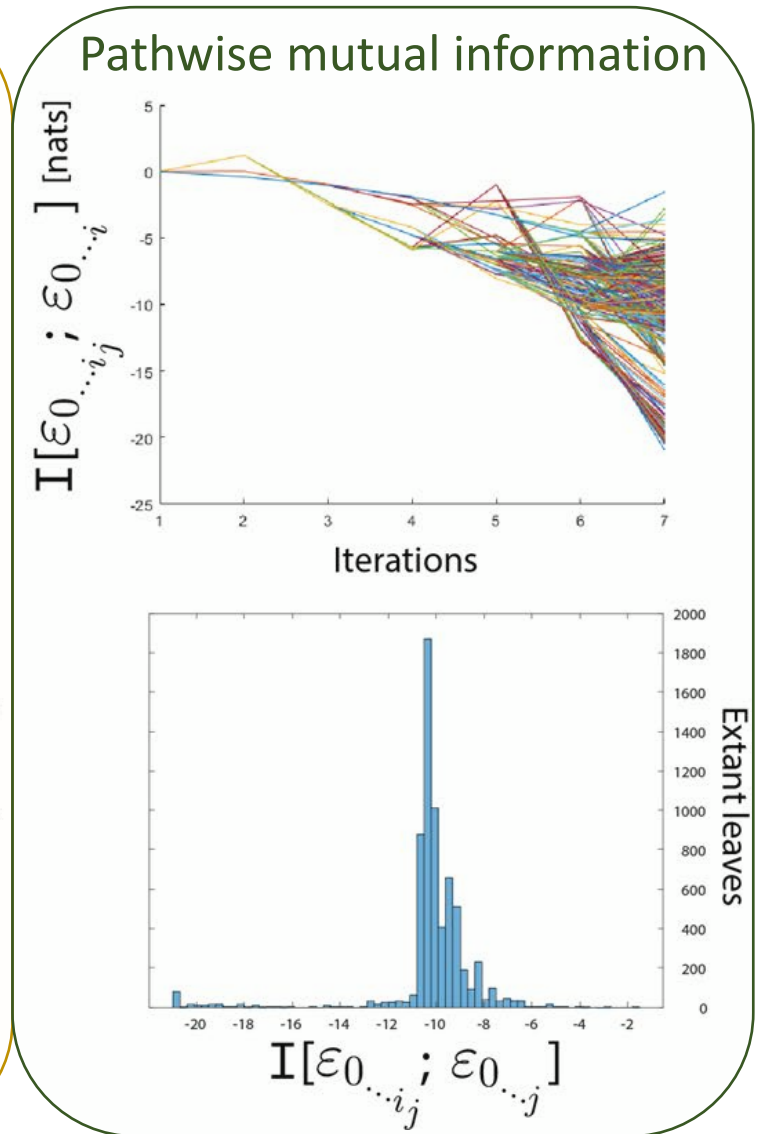
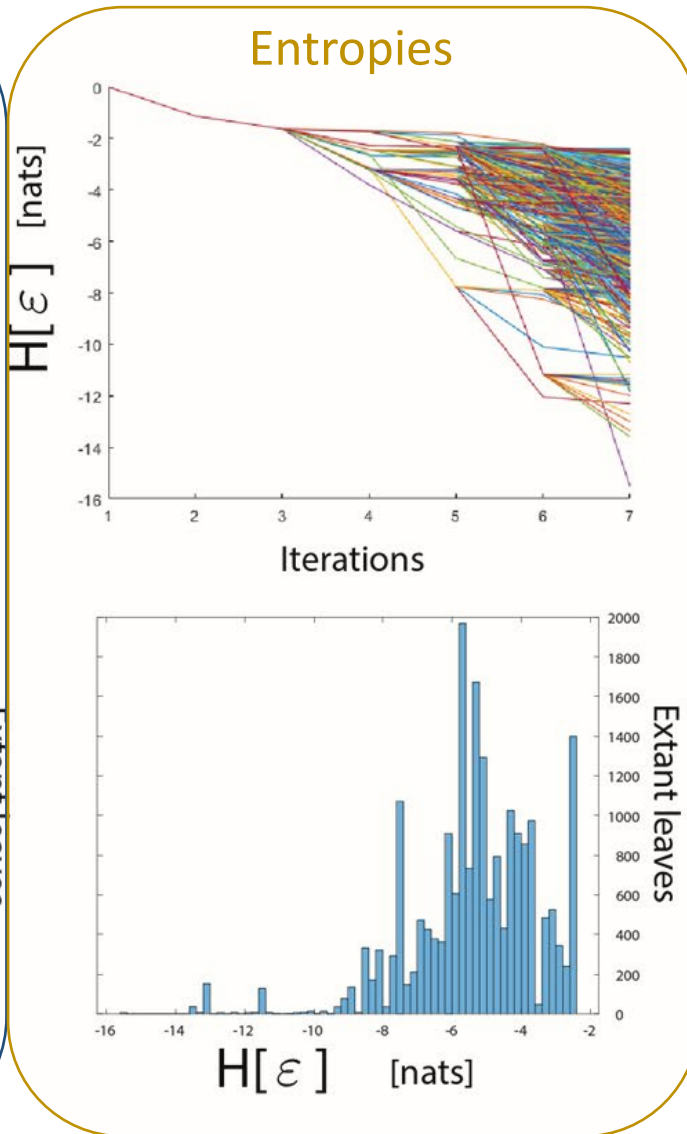
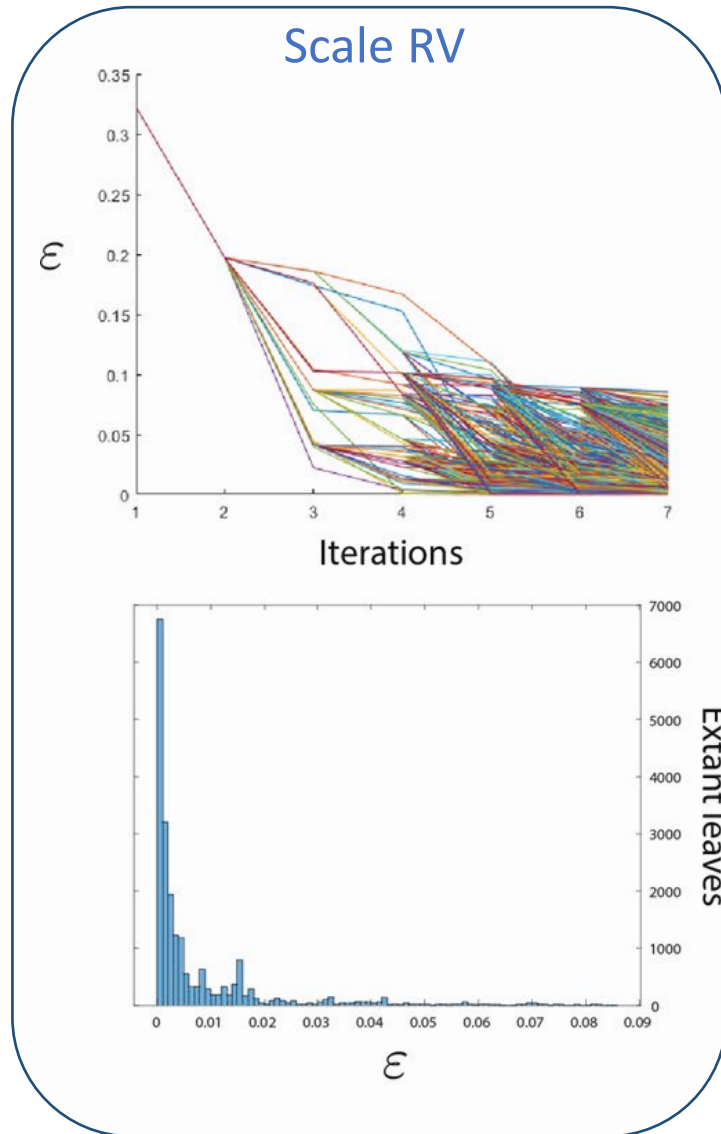
$$H[\varepsilon_{\dots i_j}] = \log(\varepsilon_{\dots i}) \leq 0$$

# Comparing information measures

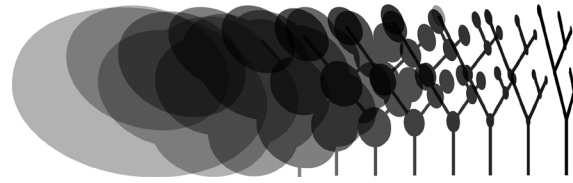
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# Example results: 7 iterations, 20643 extant leaves



# Example results: Local channel capacity densities

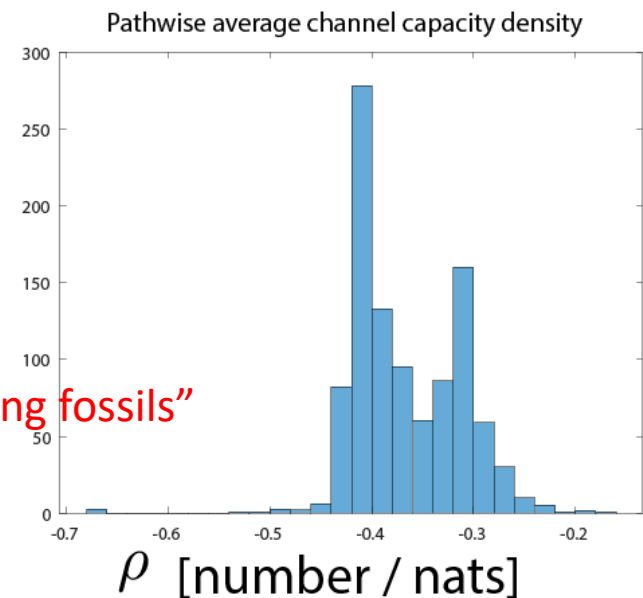
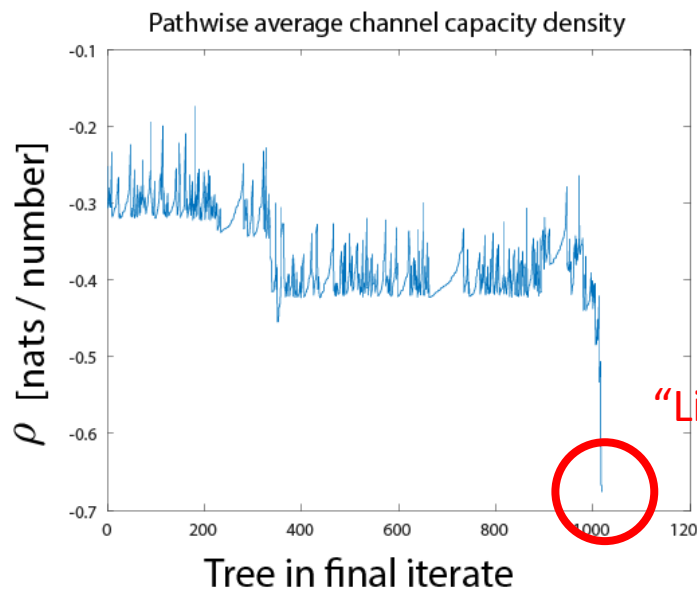
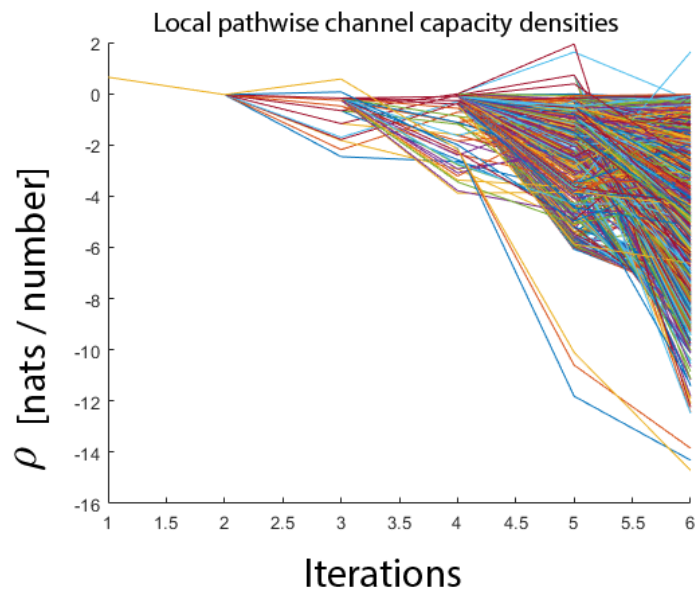


$i$  ——— Local channel ———>  $j$

Channel capacity: 
$$C_{ij} = \max_{p(\varepsilon_{0 \dots ij})} I[\varepsilon_{0 \dots ij}; \varepsilon_{0 \dots i}] = H[\varepsilon_{0 \dots ij}] + H[\varepsilon_{0 \dots i}] - H[\varepsilon_{0 \dots ij}, \varepsilon_{0 \dots i}] = \log(\varepsilon_{0 \dots ij}) + \log(\varepsilon_{0 \dots i}) - \log\left(\frac{\varepsilon_{0 \dots ij}}{\varepsilon_{0 \dots i}}\right)$$

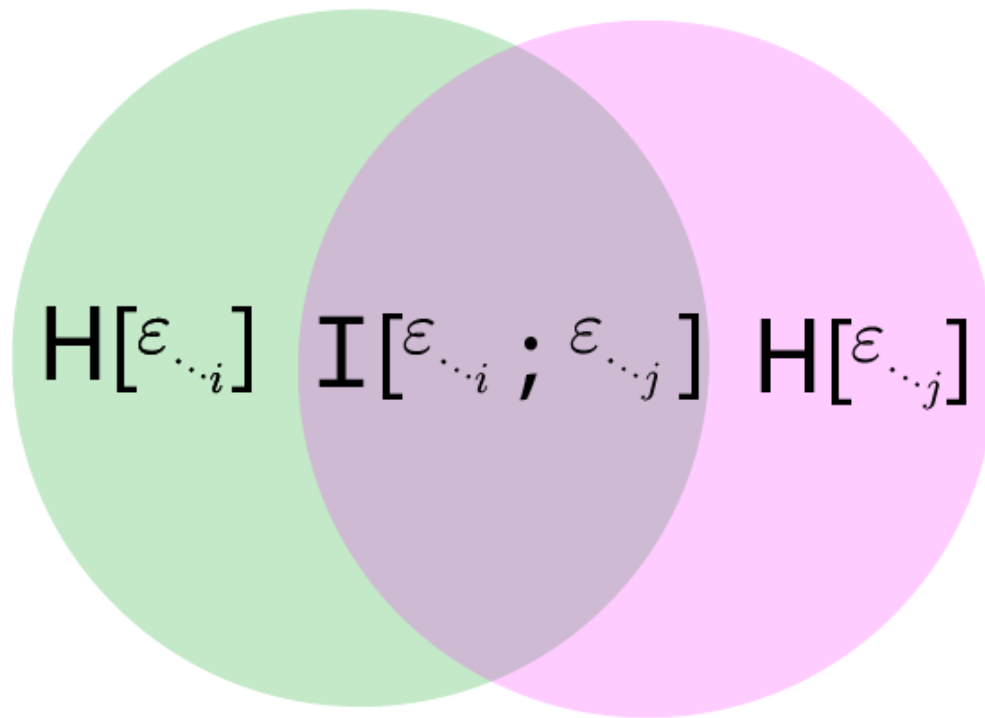
Channel capacity density: 
$$\rho_{ij} = \frac{C_{ij}}{L(\mathcal{T}_{ij})}$$

7 iterations, 4,430 extant leaves, 1,018 trees in 7<sup>th</sup> iterate

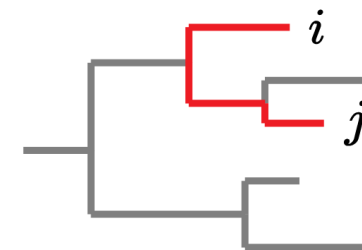


# Mutual information and information distance:

How do extant leaves relate?



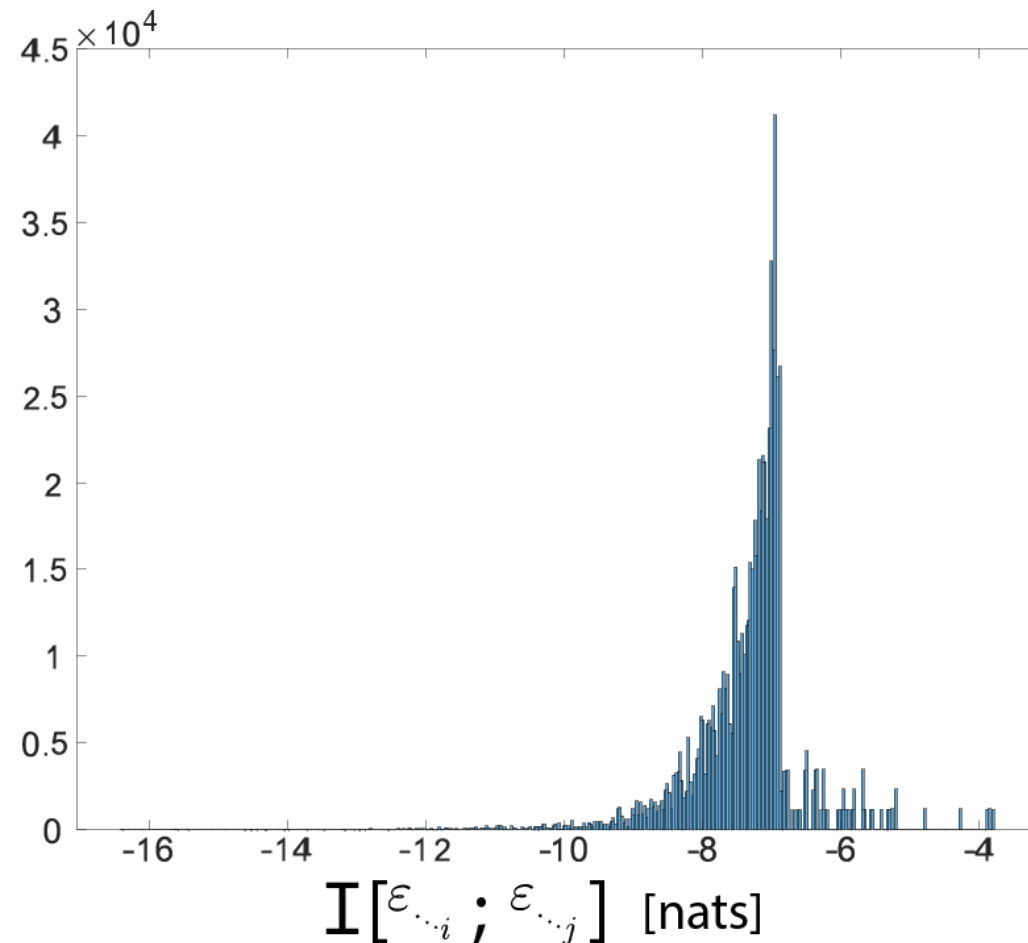
$$d_{\varepsilon}(\varepsilon_{\dots i} ; \varepsilon_{\dots j}) := \Delta\varepsilon_{ij} \text{ [form]}$$



Assign **information distance**  $d_{\varepsilon}$  to be **the metric of biological form**.

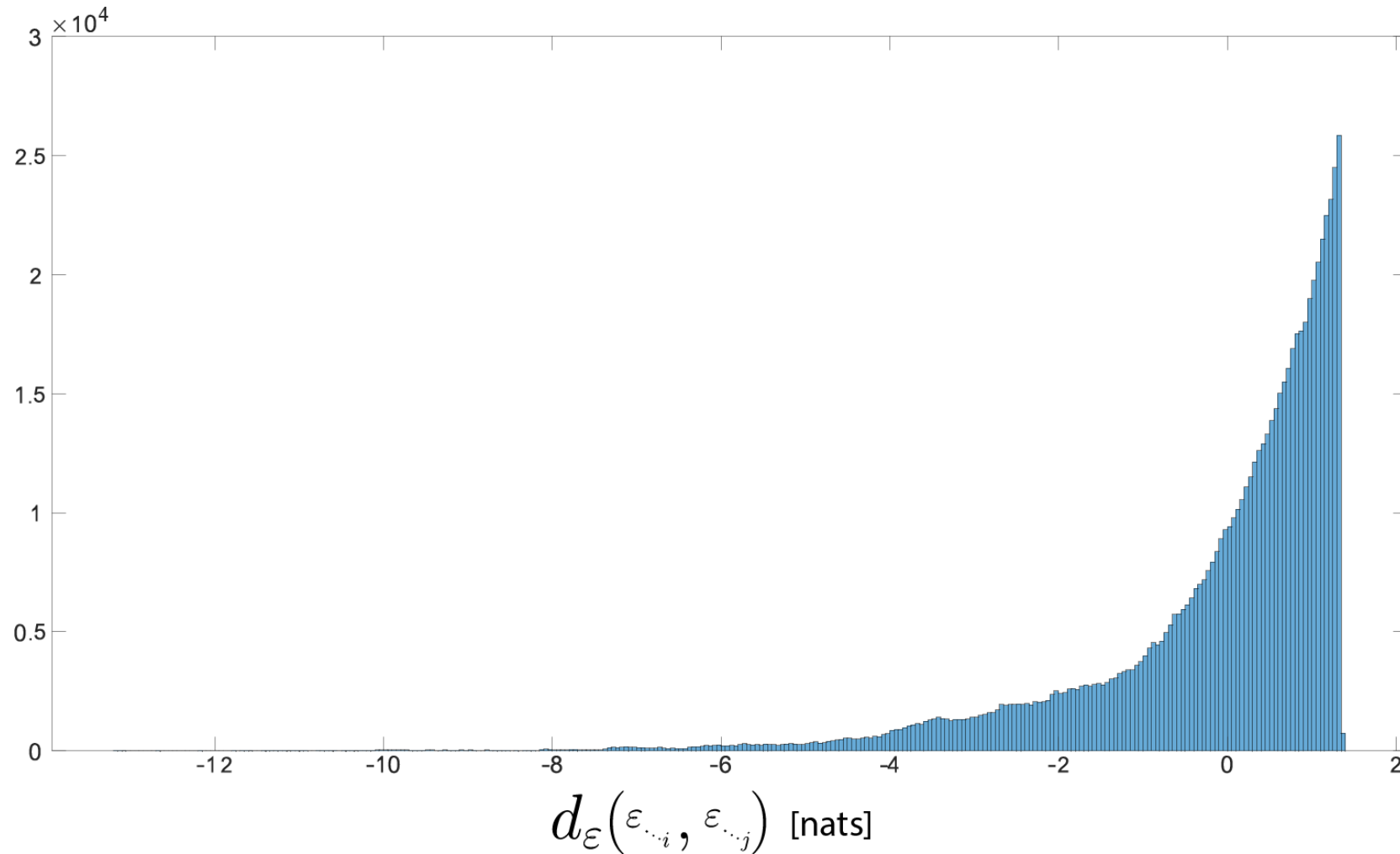
# Example results: cross-path mutual information

All unique pairwise combinations of mutual information for 6,616 extant leaves (7 iterations)



# Example results: cross-path information distance

All unique pairwise combinations of information distance for 6,616 extant leaves (7 iterations)

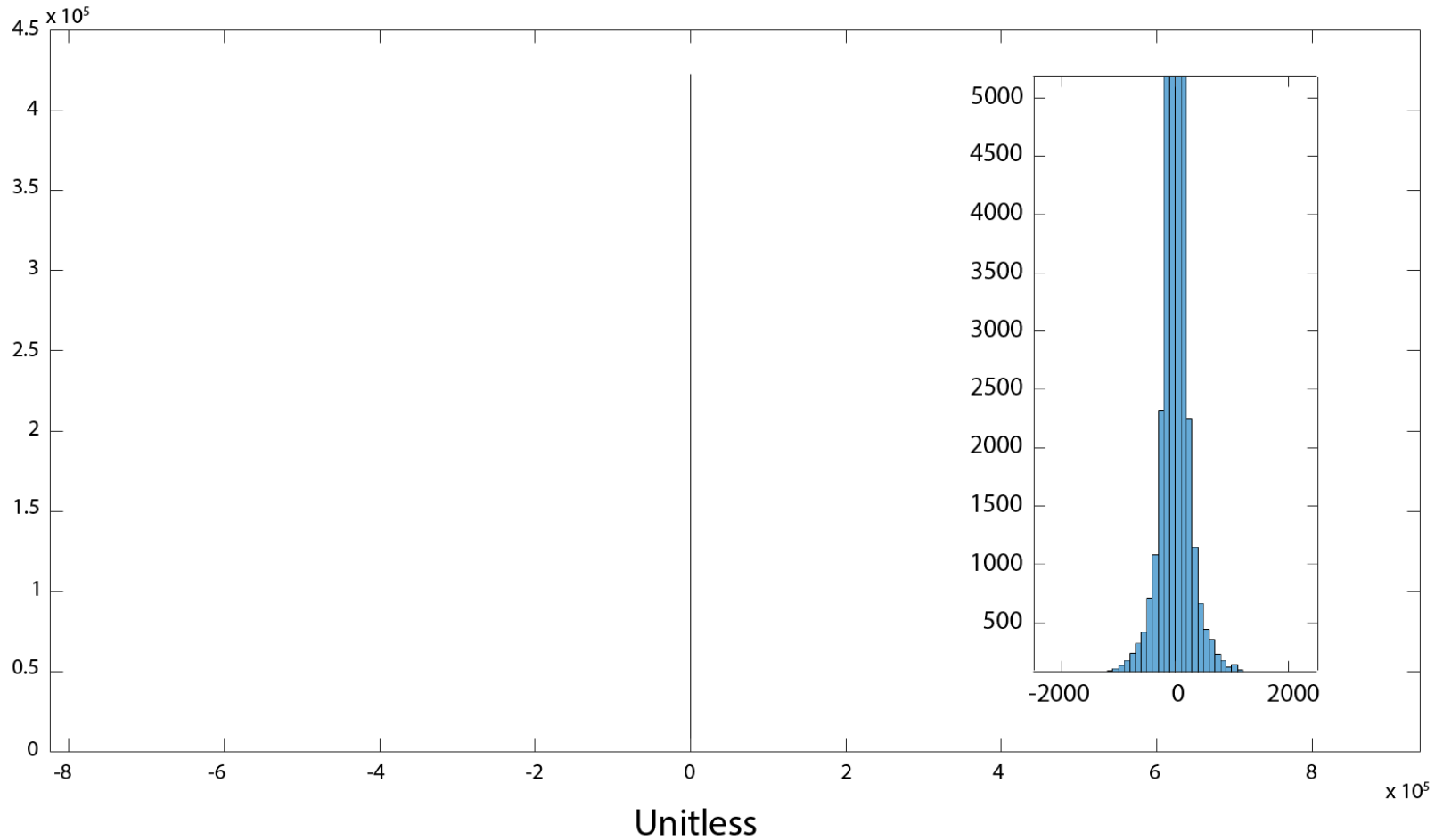




Interested in:

$$\frac{I[\varepsilon_{\dots i}, \varepsilon_{\dots j}]}{d_{\varepsilon}(\varepsilon_{\dots i}, \varepsilon_{\dots j})}$$

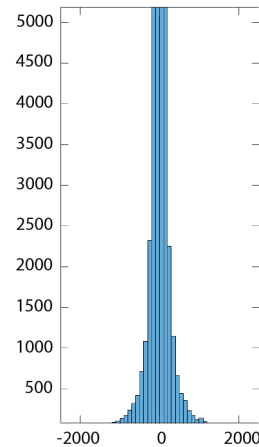
All unique pairwise combinations of  $\frac{I}{d_{\varepsilon}}$  for 6,616 extant leaves (7 iterations)



The metric  $d_t(\varepsilon_{\cdot\cdot i}, \varepsilon_{\cdot\cdot j})$ ? For another day

Multifractal structure  $\Rightarrow$  not a simple task

Need  $\frac{I[\varepsilon_{\cdot\cdot i}, \varepsilon_{\cdot\cdot j}]}{d_\varepsilon(\varepsilon_{\cdot\cdot i}, \varepsilon_{\cdot\cdot j})}$  and the multifractal spectrum





Thanks