

Machine learning quantification of cluster structure

Oliver Lomax

Matthew Bates & Anthony Whitworth

6th September 2018

The Wonders of Star Formation

Outline

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- Q parameter, used to measure substructure in star clusters, is unreliable.

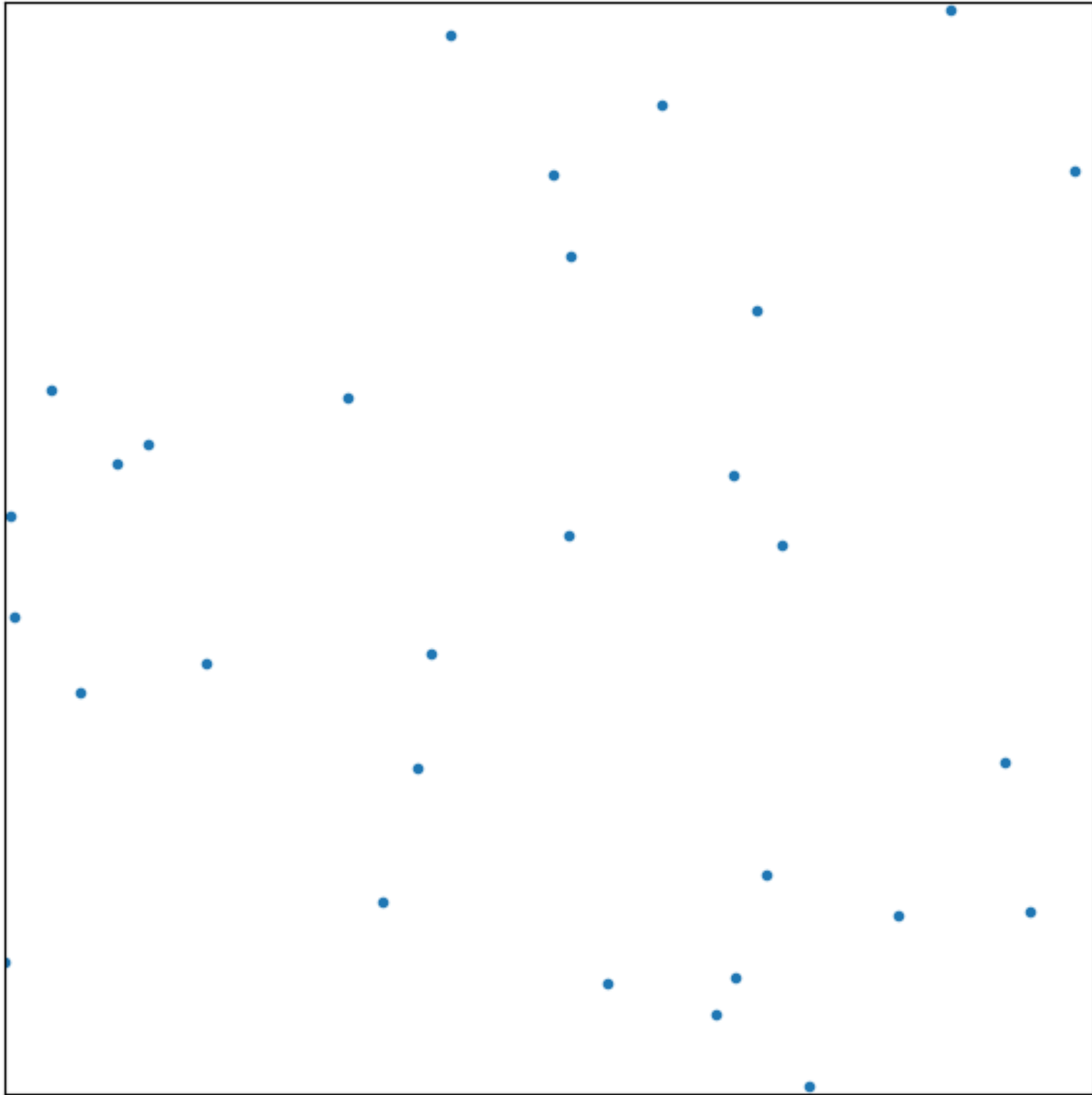
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- We have tools available to do better!

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- Disclaimer: I am neither an observer or an N -body numericist!

Q parameter

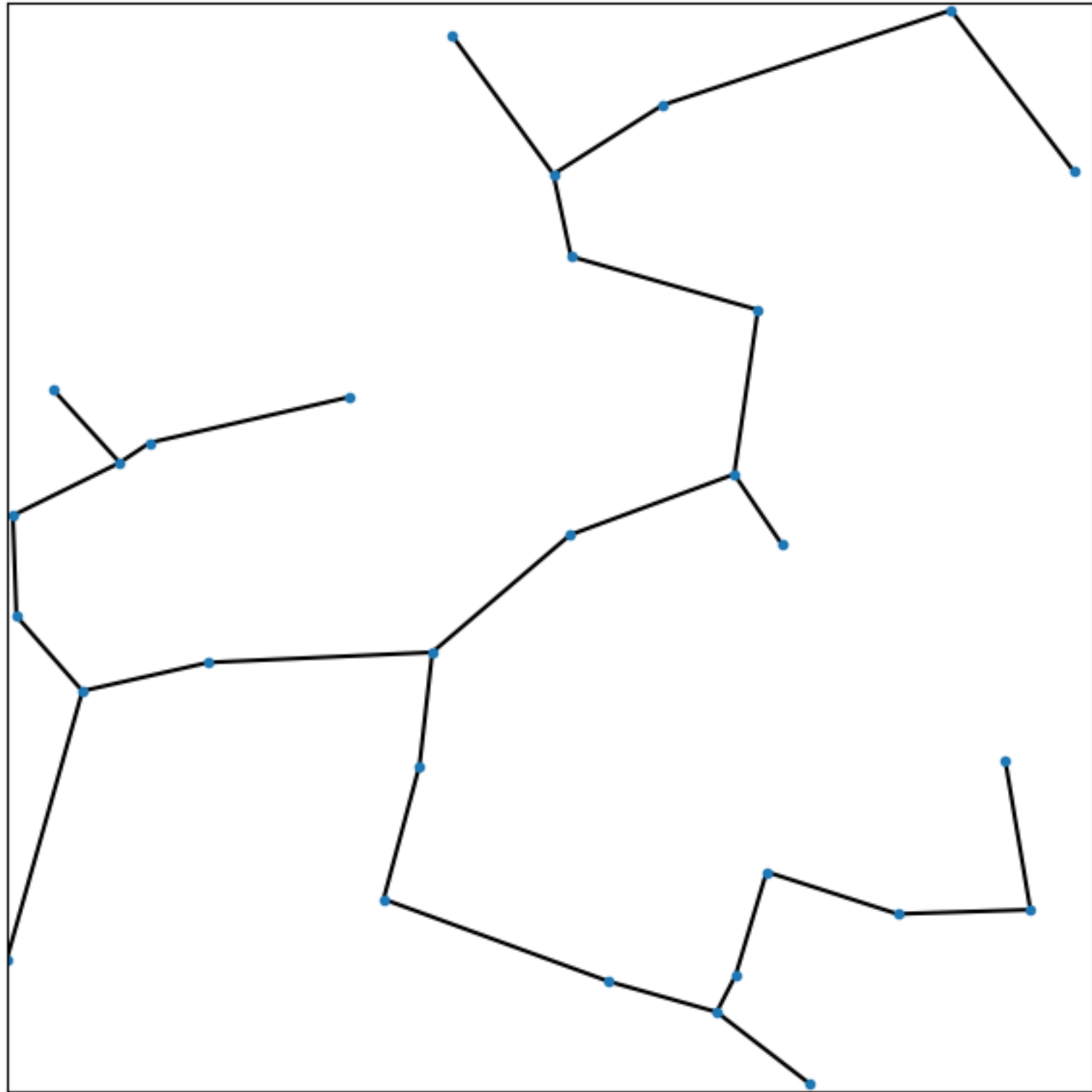
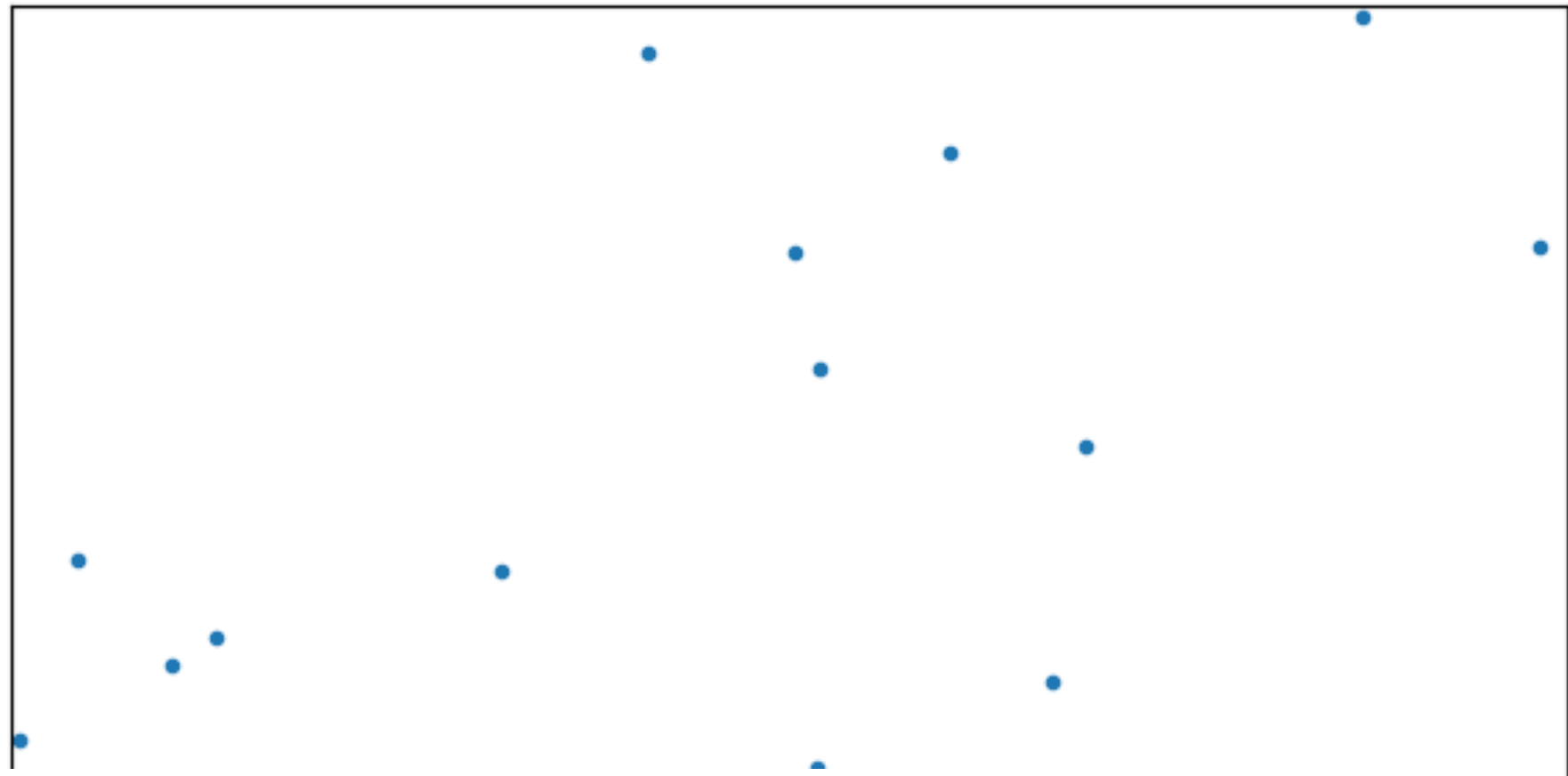


$$Q = \frac{\bar{m}}{\bar{s}}$$

$$\bar{m} = \frac{\sum_{i=1}^{N_m} m_i}{(N_m + 1)^{\frac{E-1}{E}} R}$$

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Q parameter



Minimum Spanning Tree

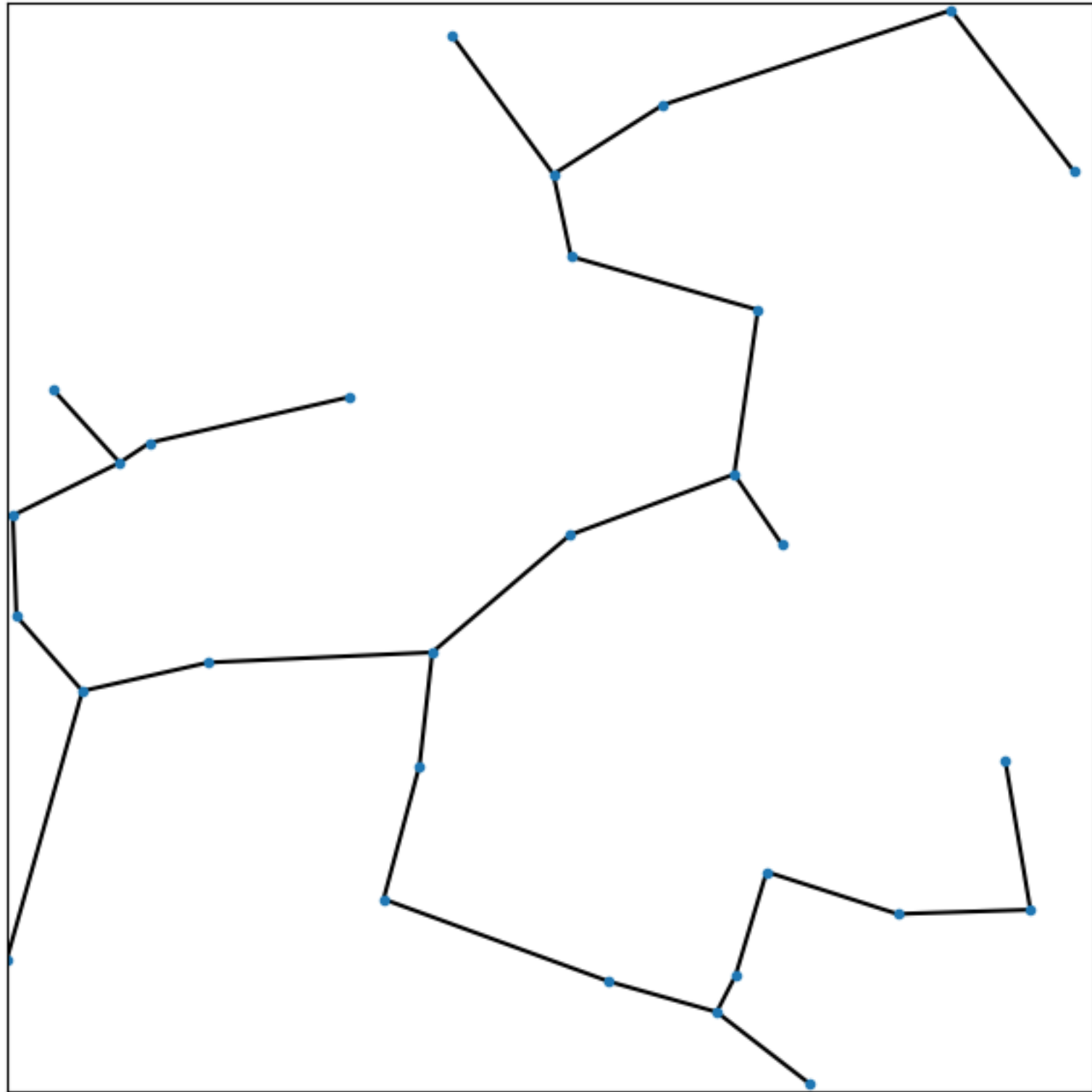
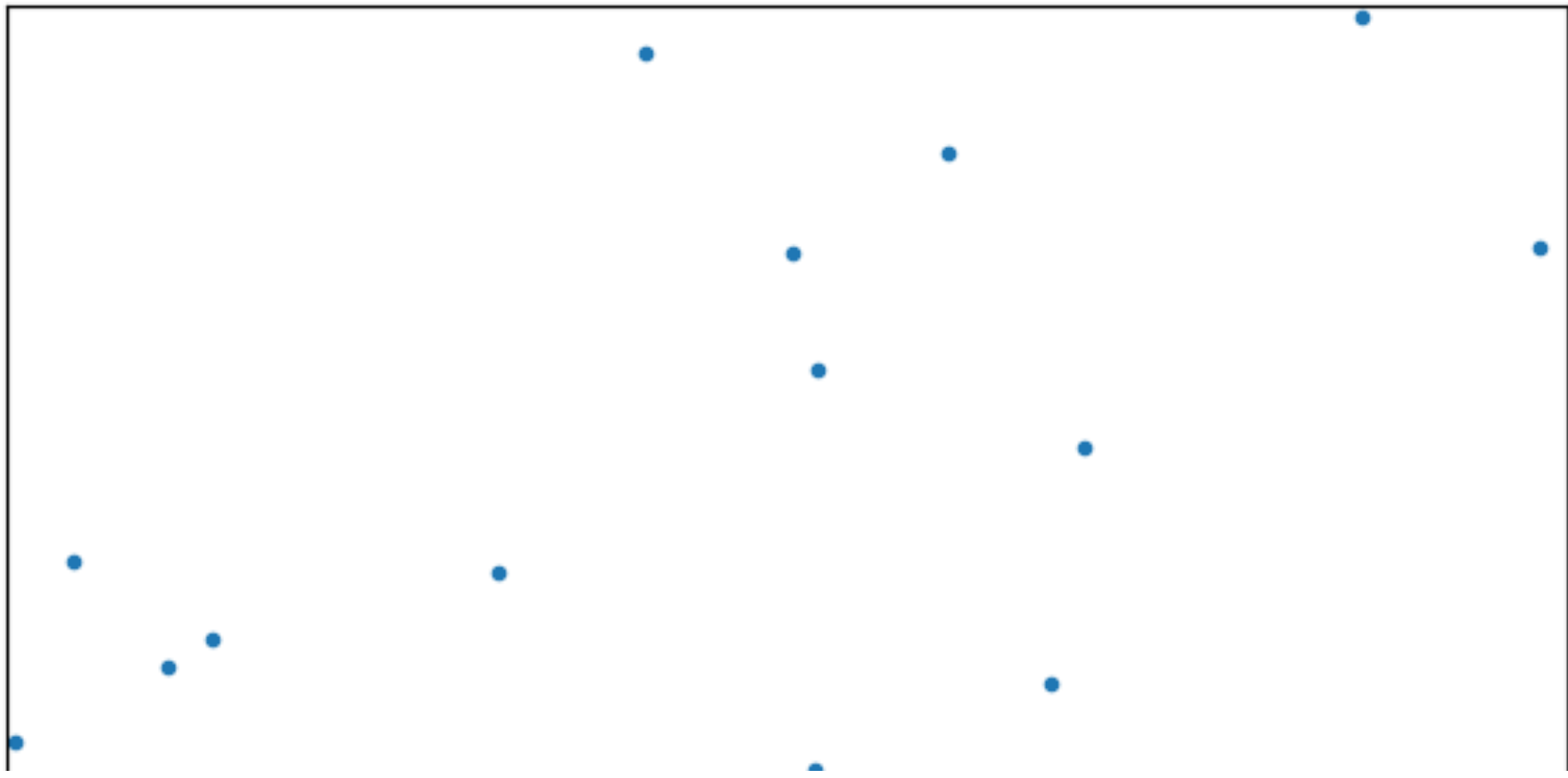


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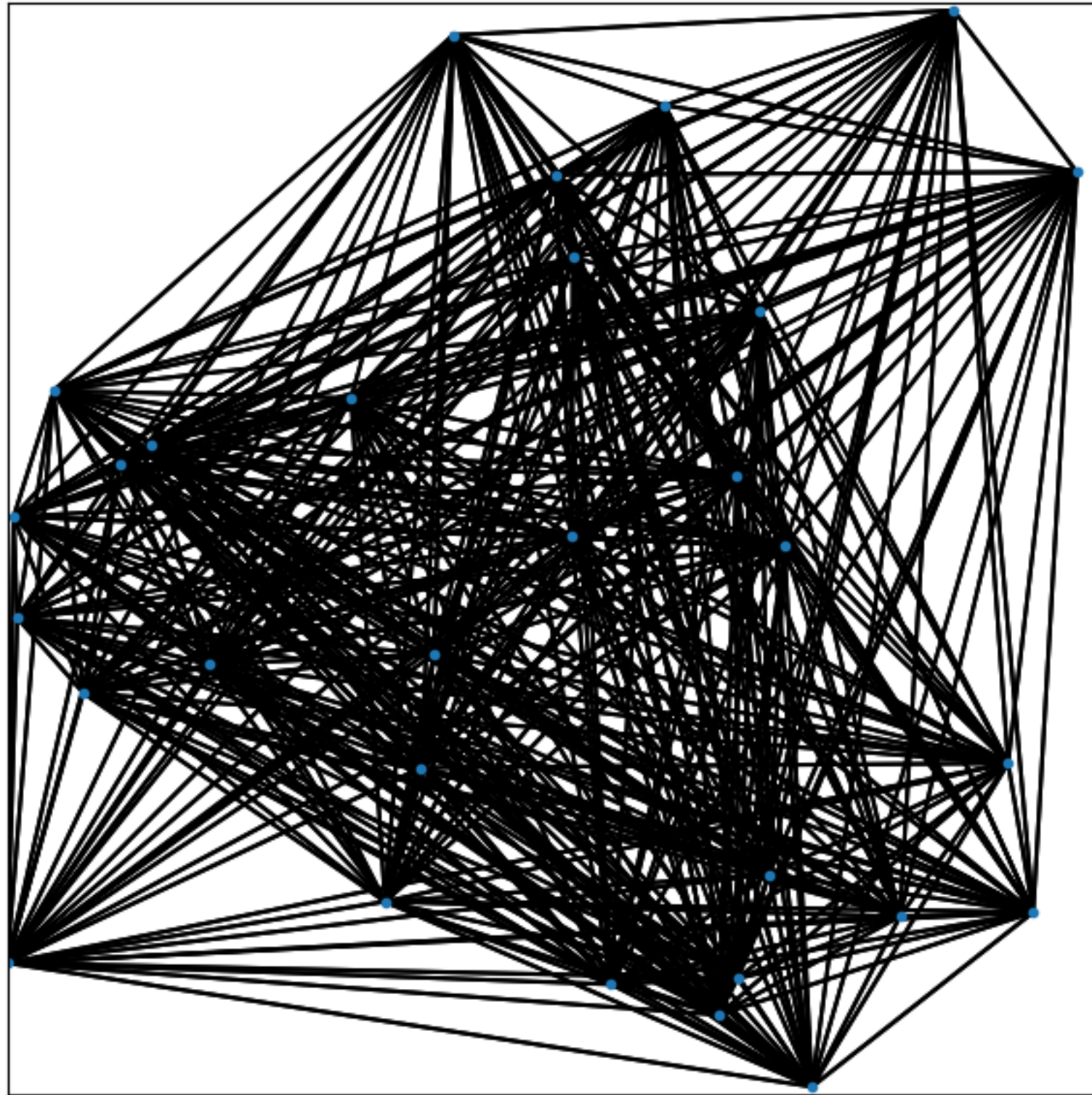
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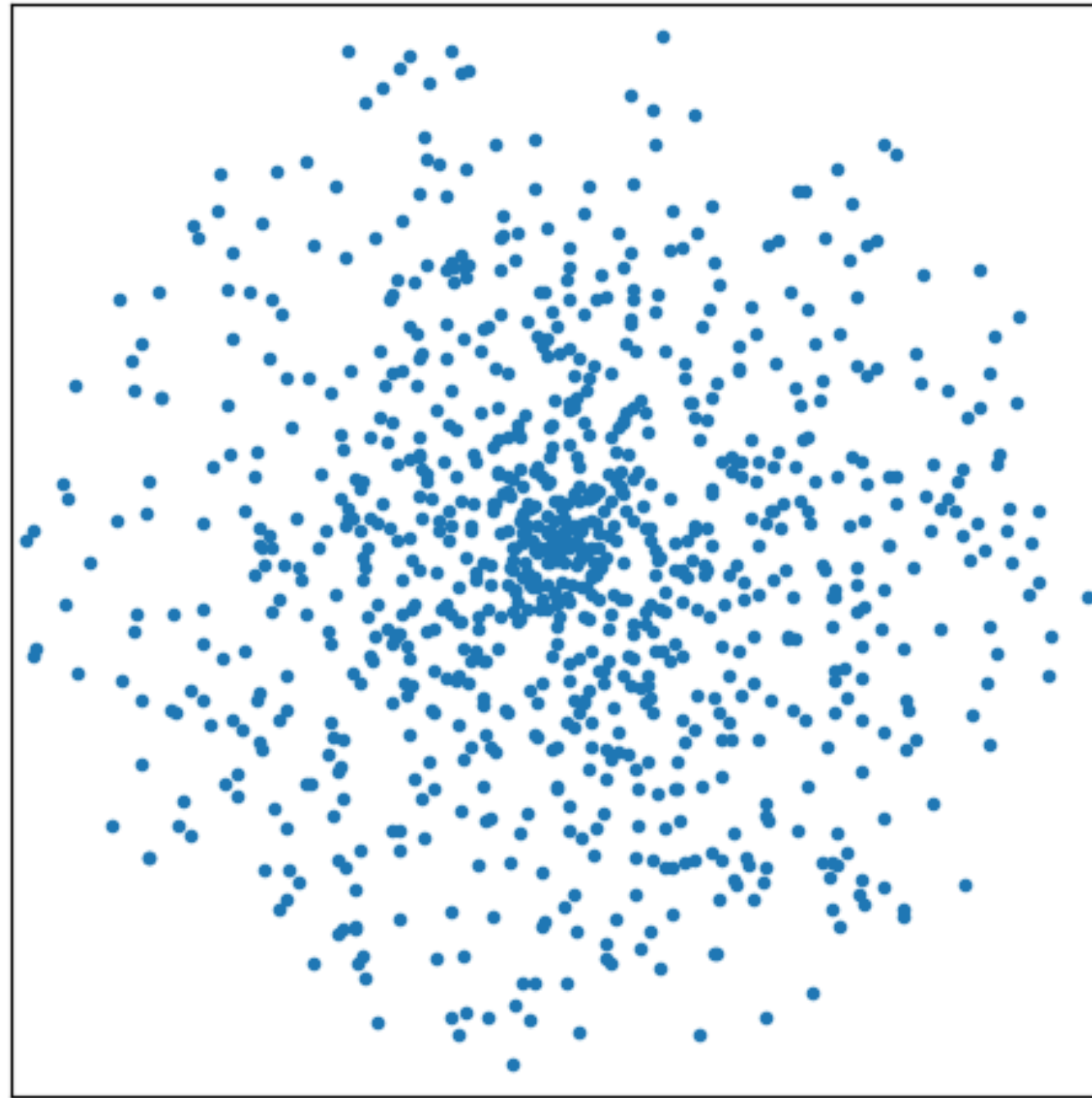
Complete Graph

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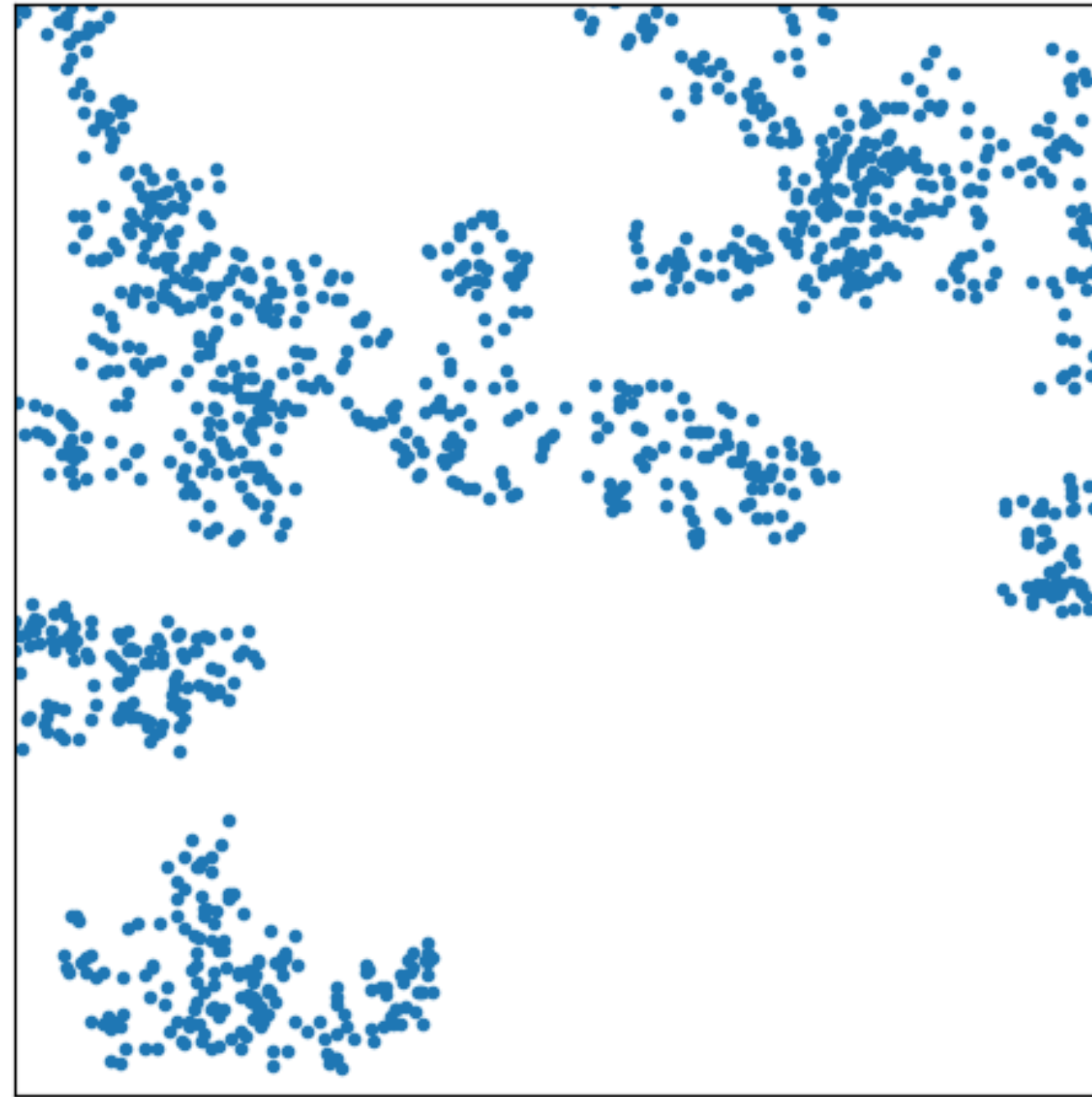
Q parameter



Radial Profile

$$\rho(r) \propto r^{-\alpha}$$

$$Q \gtrsim 0.8$$



Box Fractal

$$P(\text{sub-cube}) = \frac{2^D}{2^E}$$

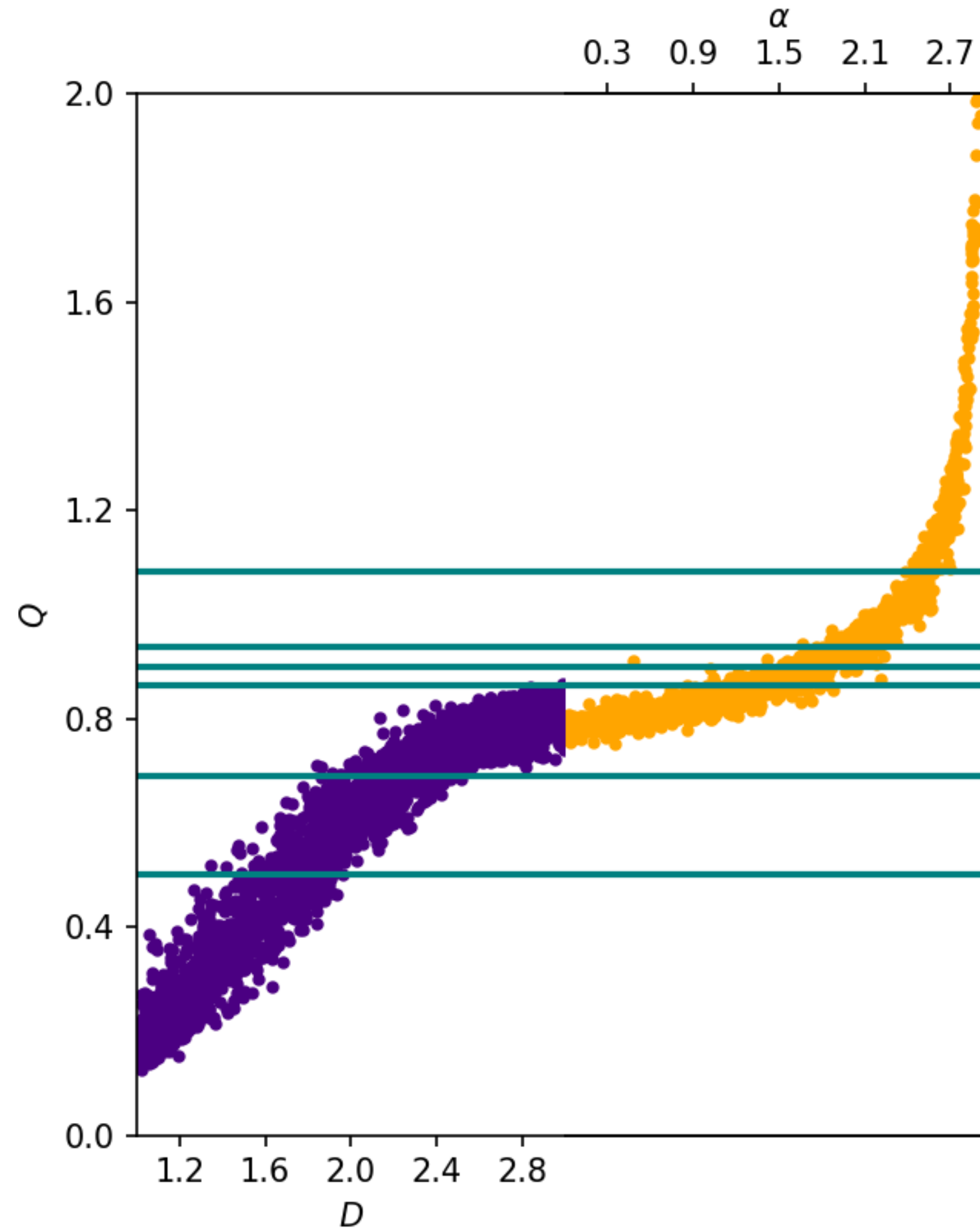
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Q parameter

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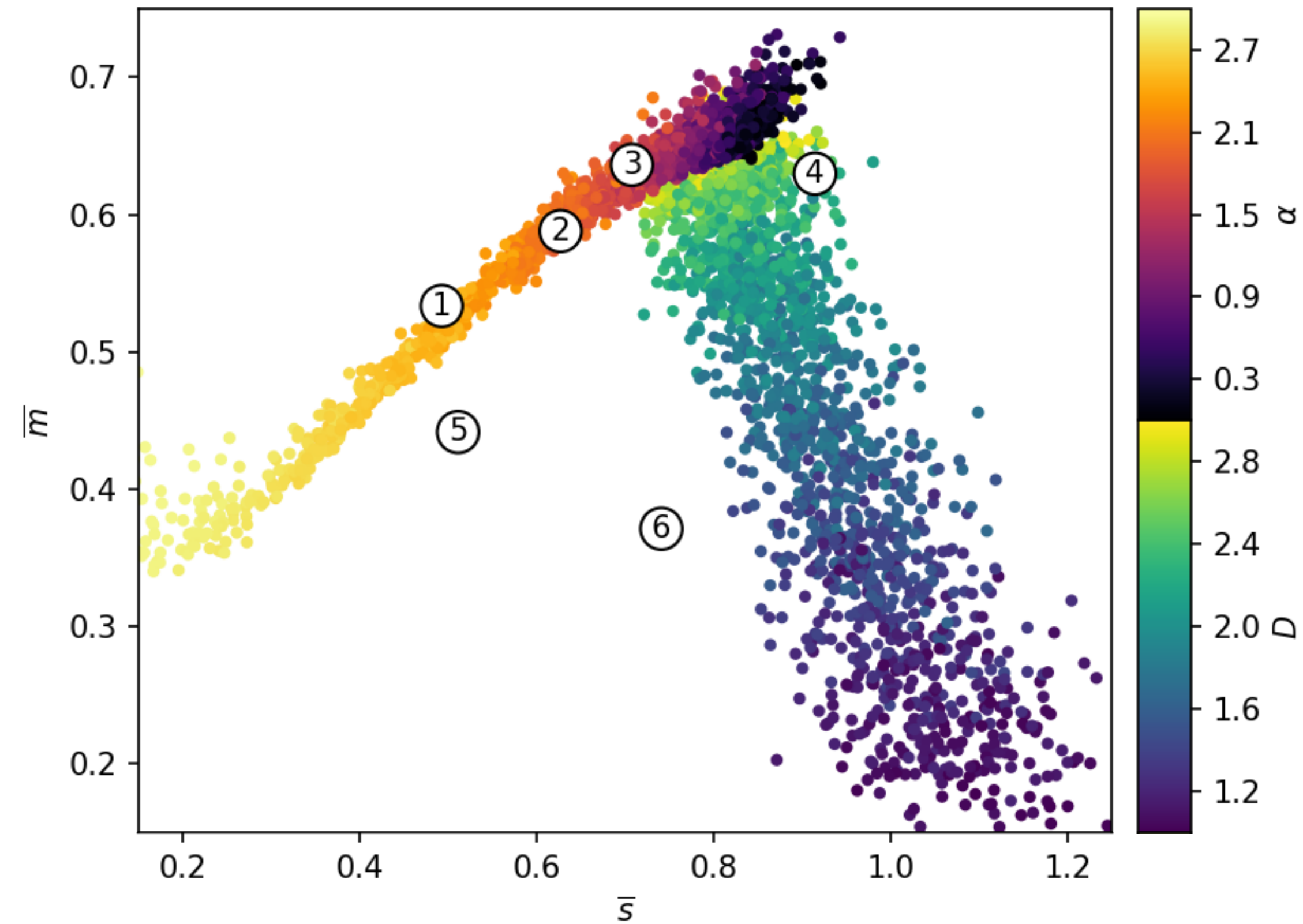
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**Cartwright and Whitworth, 2004,
 MNRAS, 348, 589**

$\bar{m}-\bar{s}$ plots

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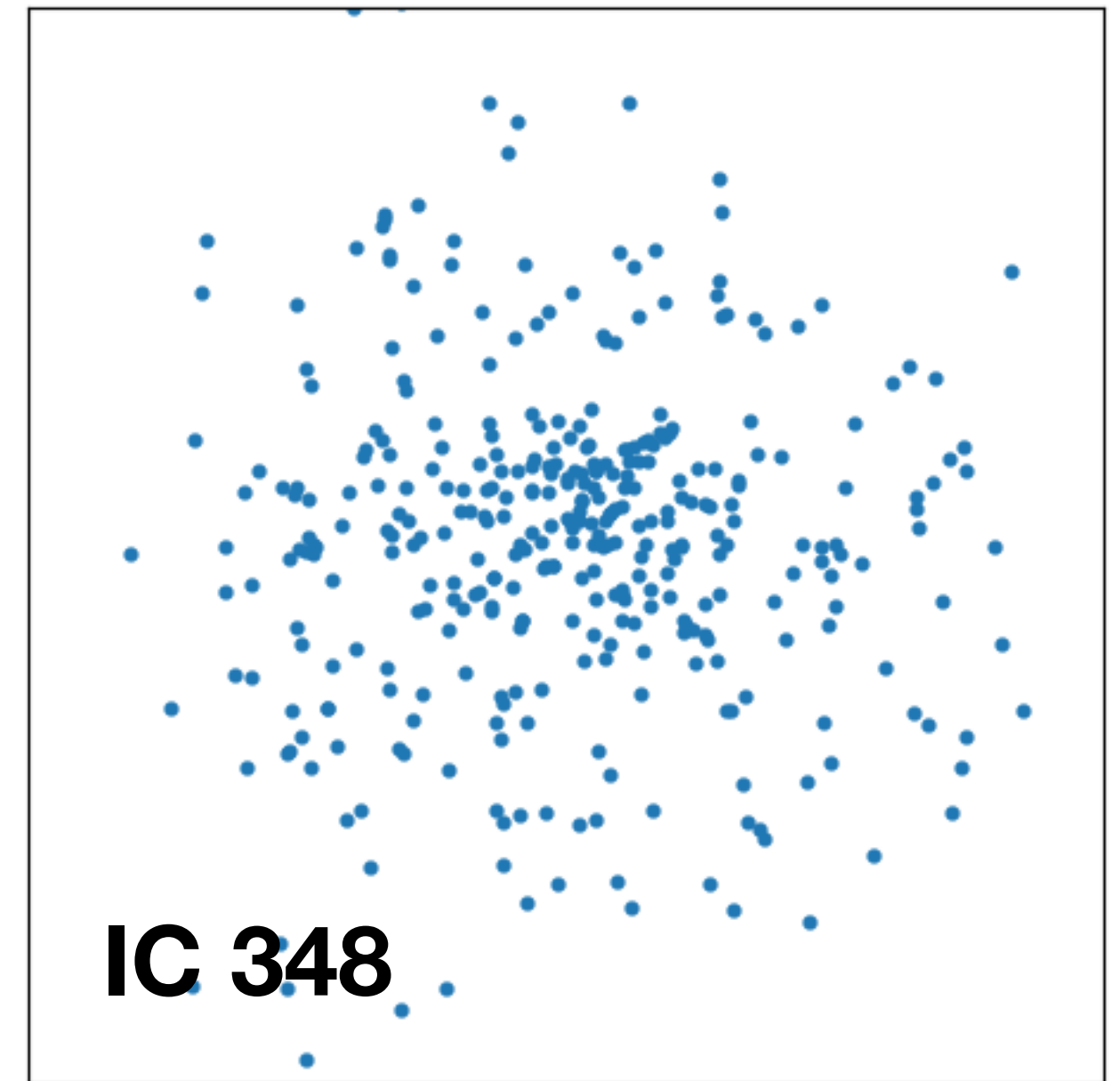
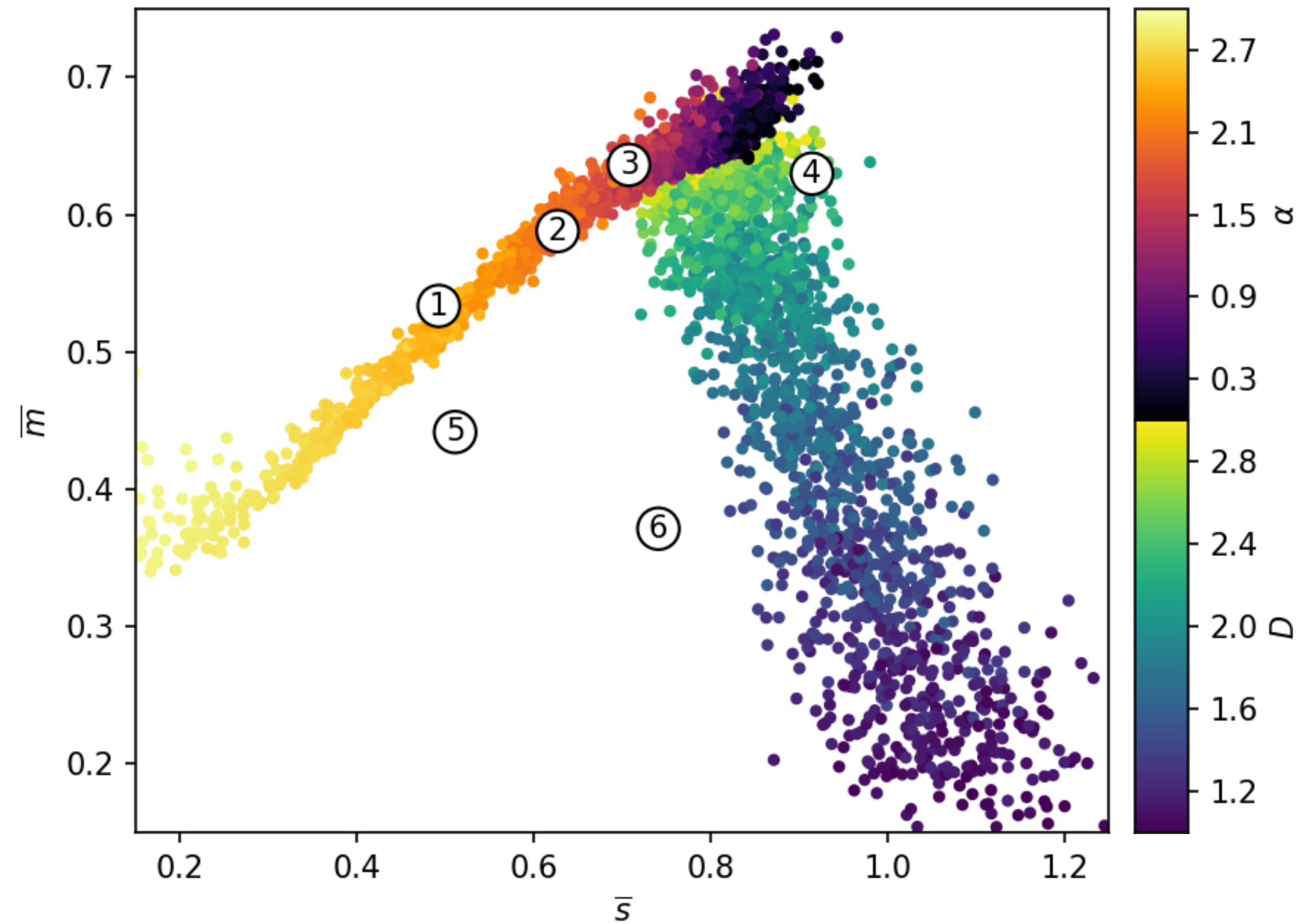


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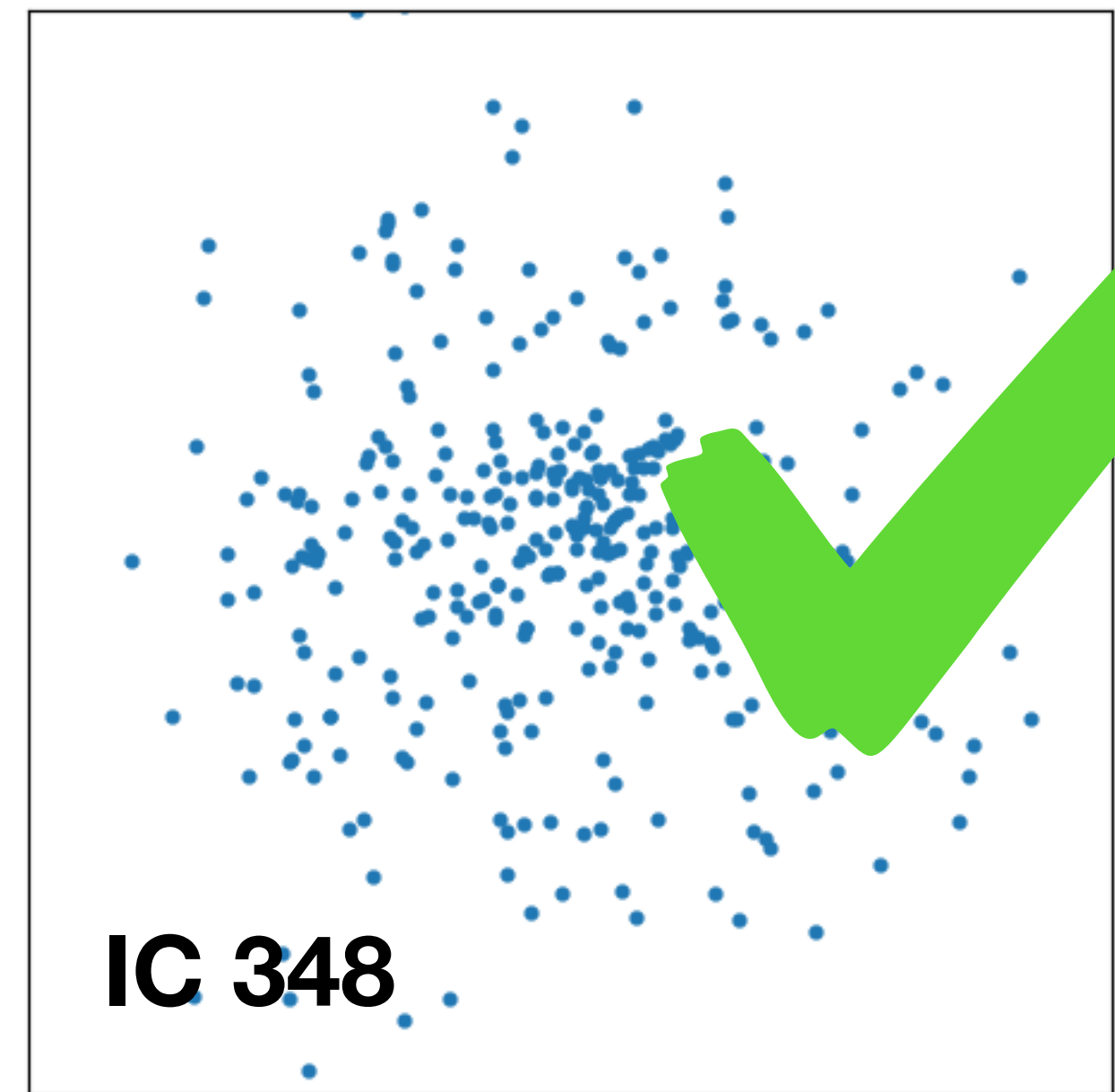
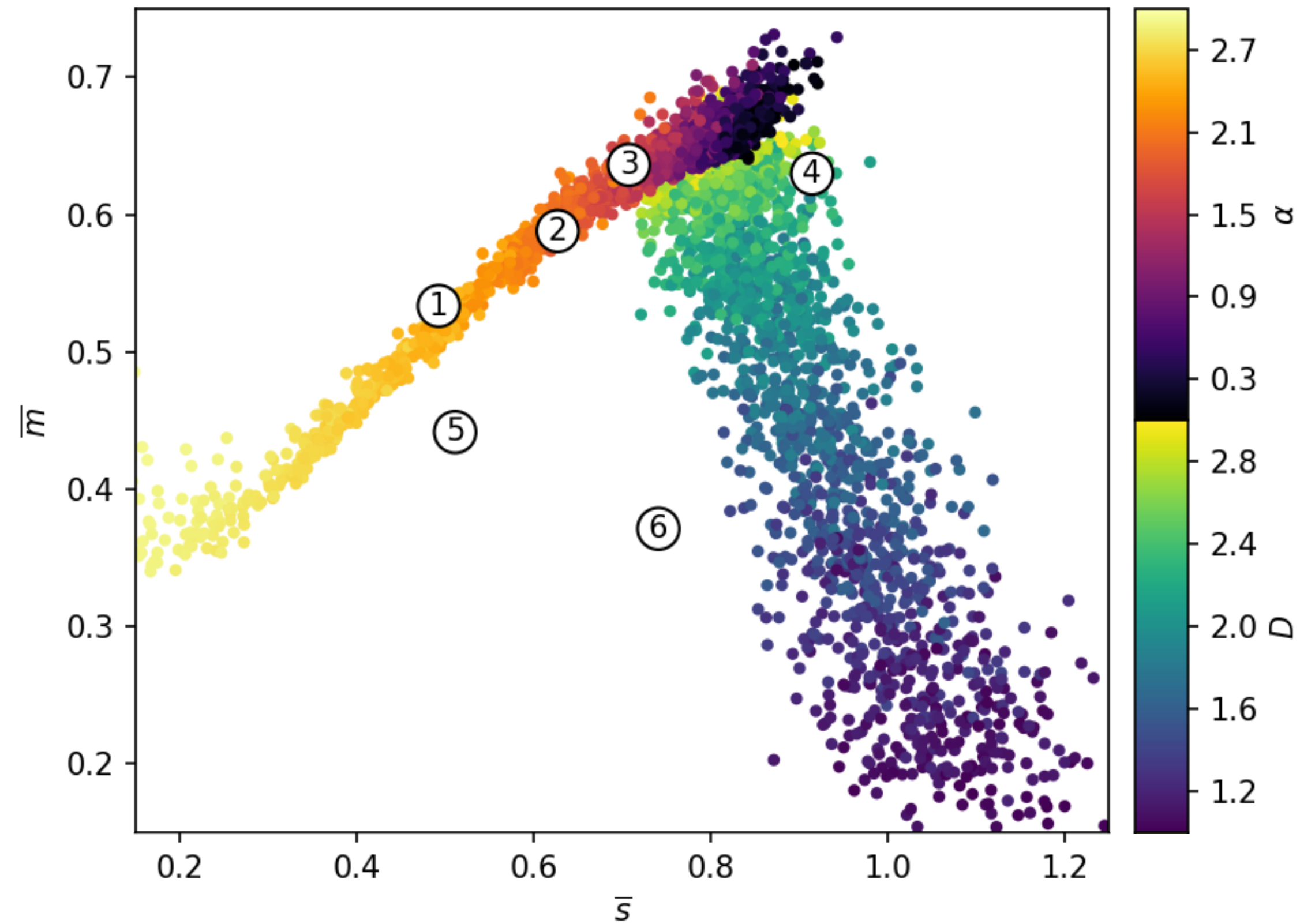


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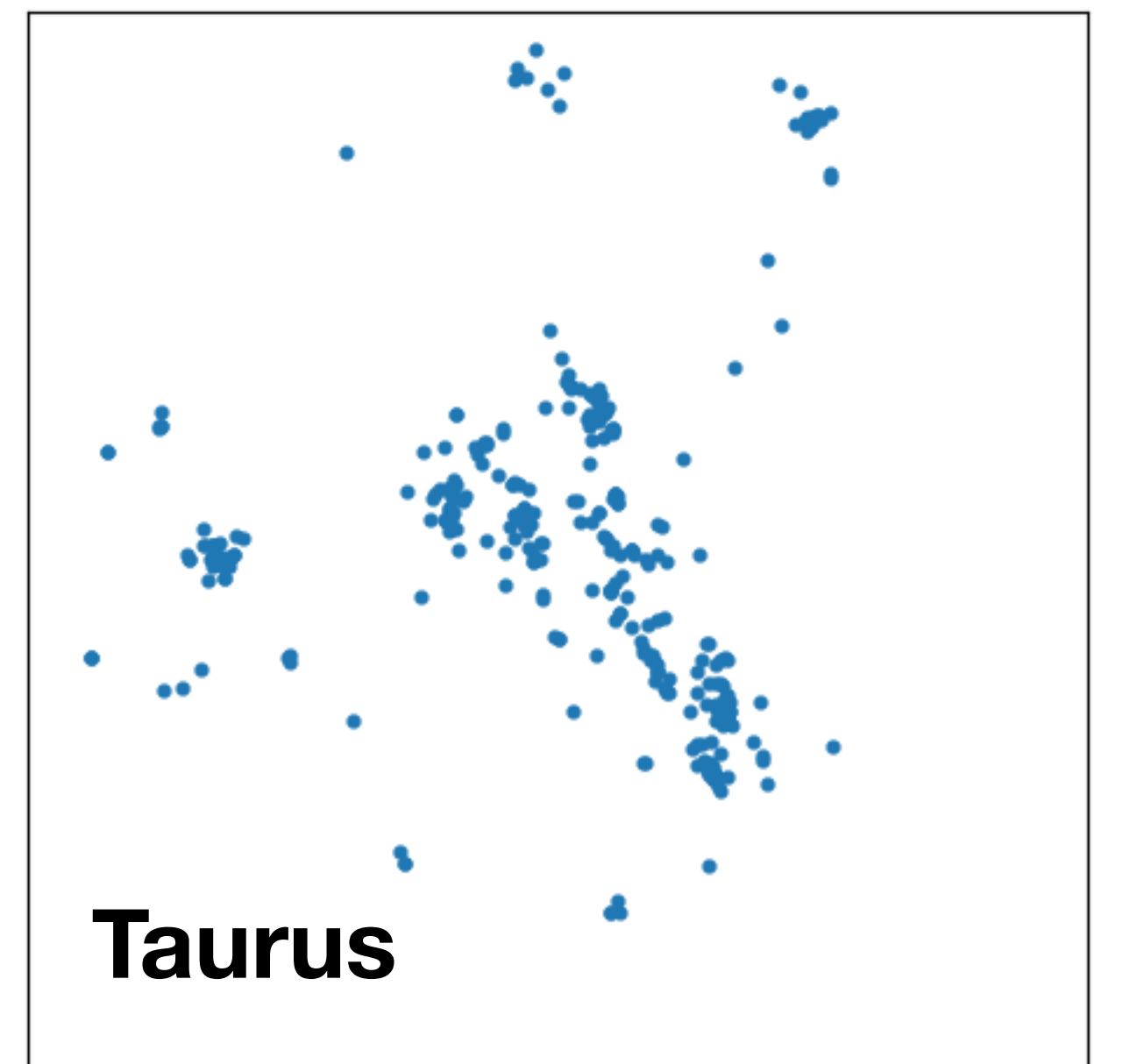
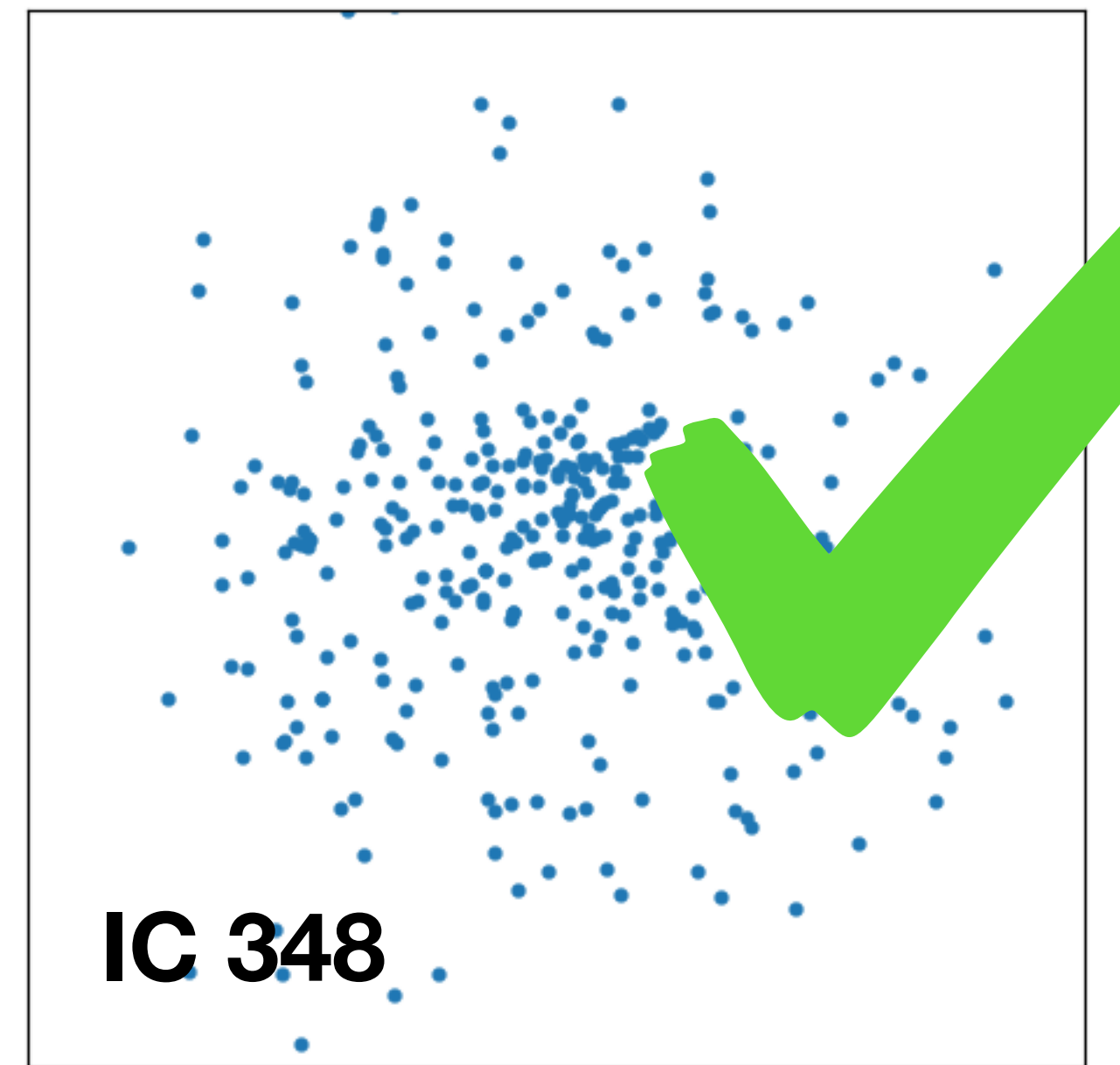
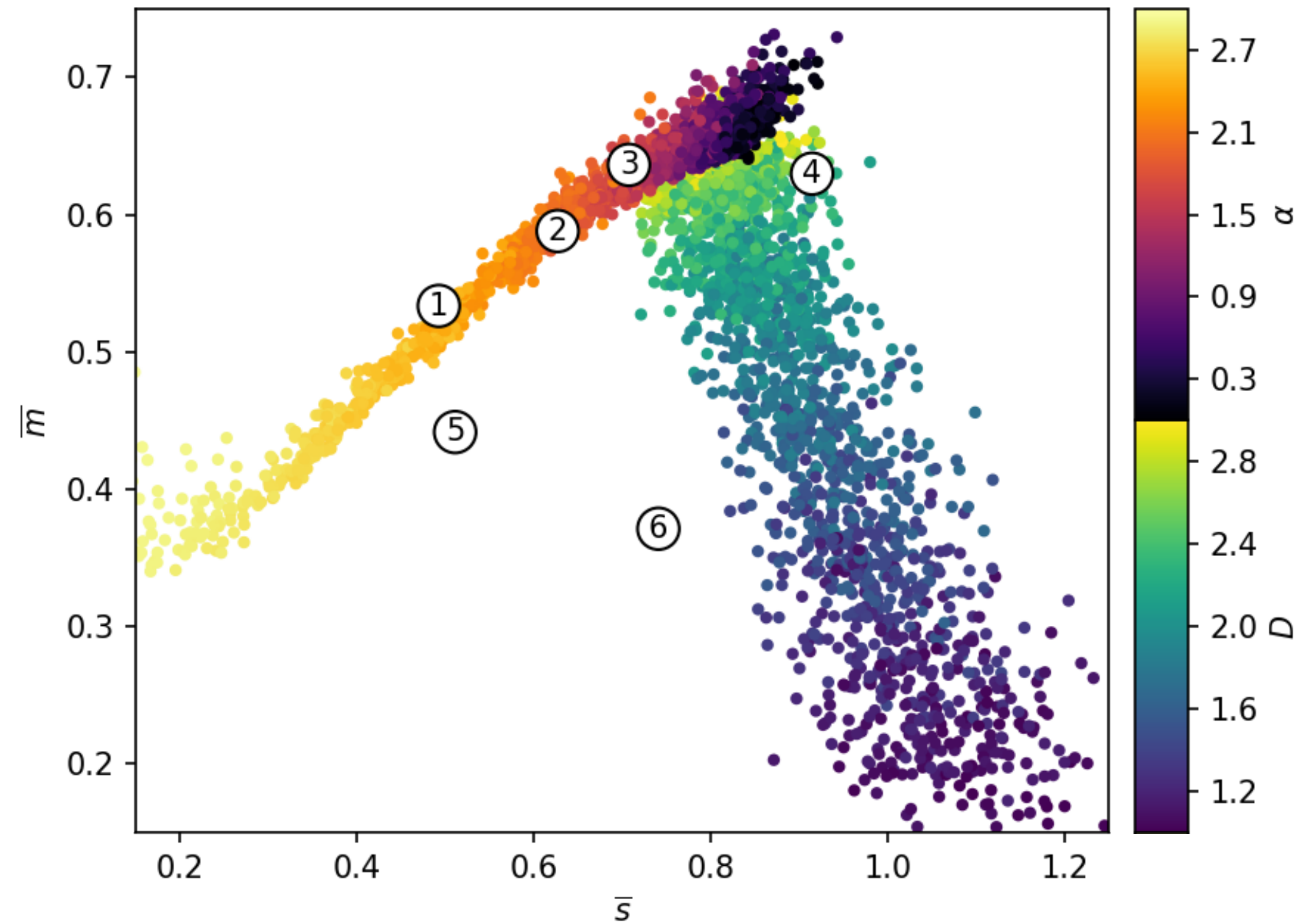


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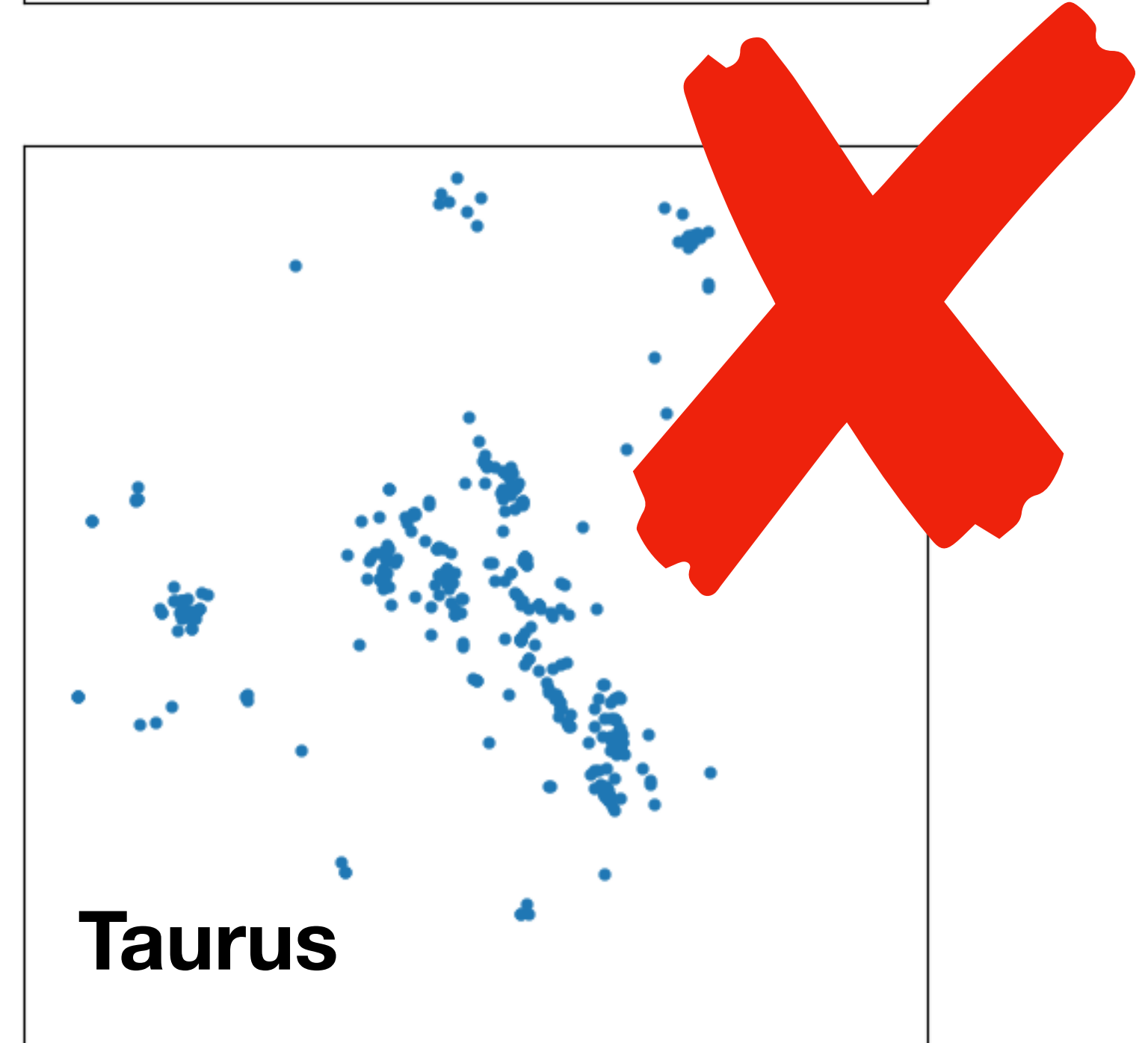
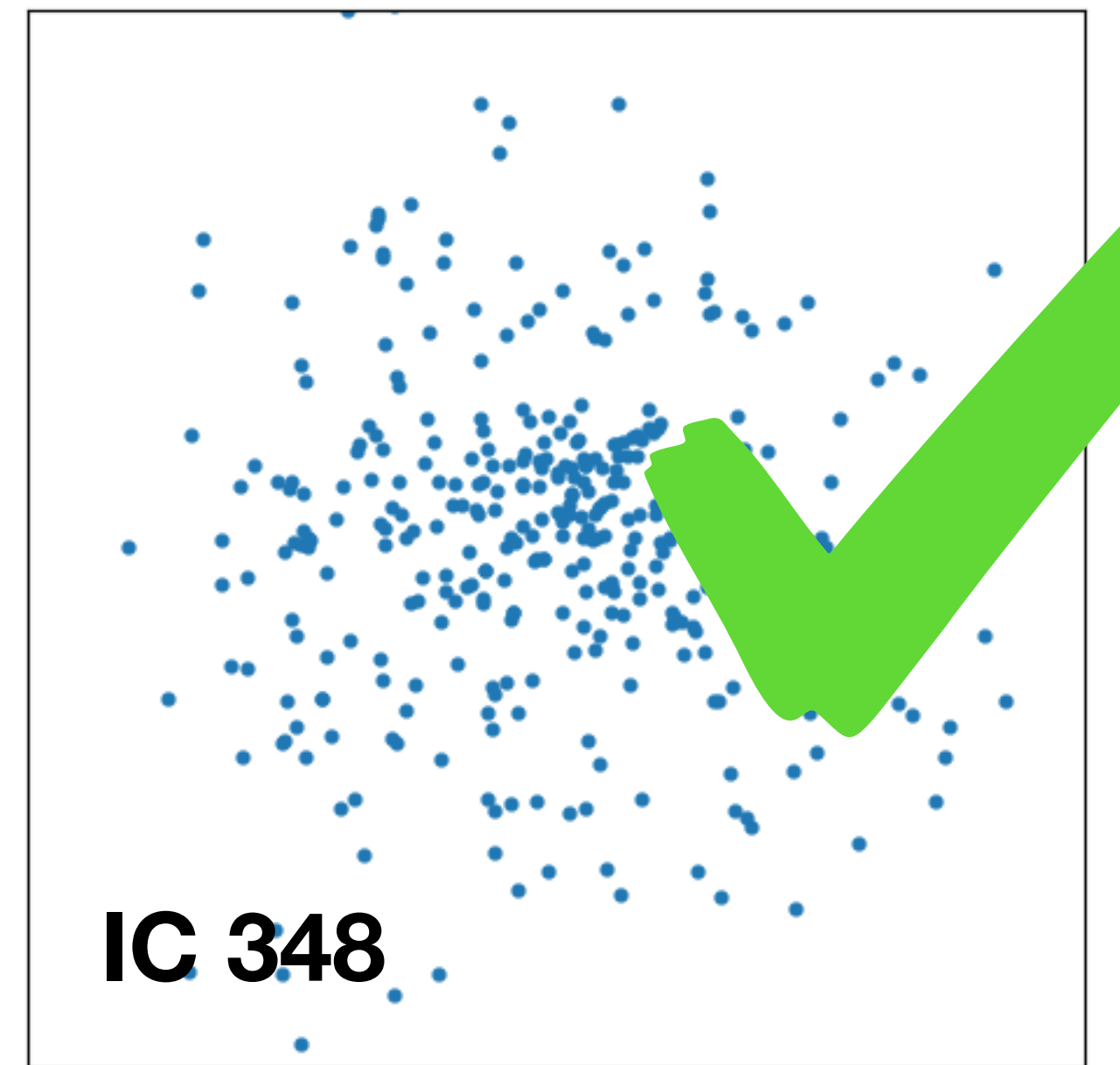
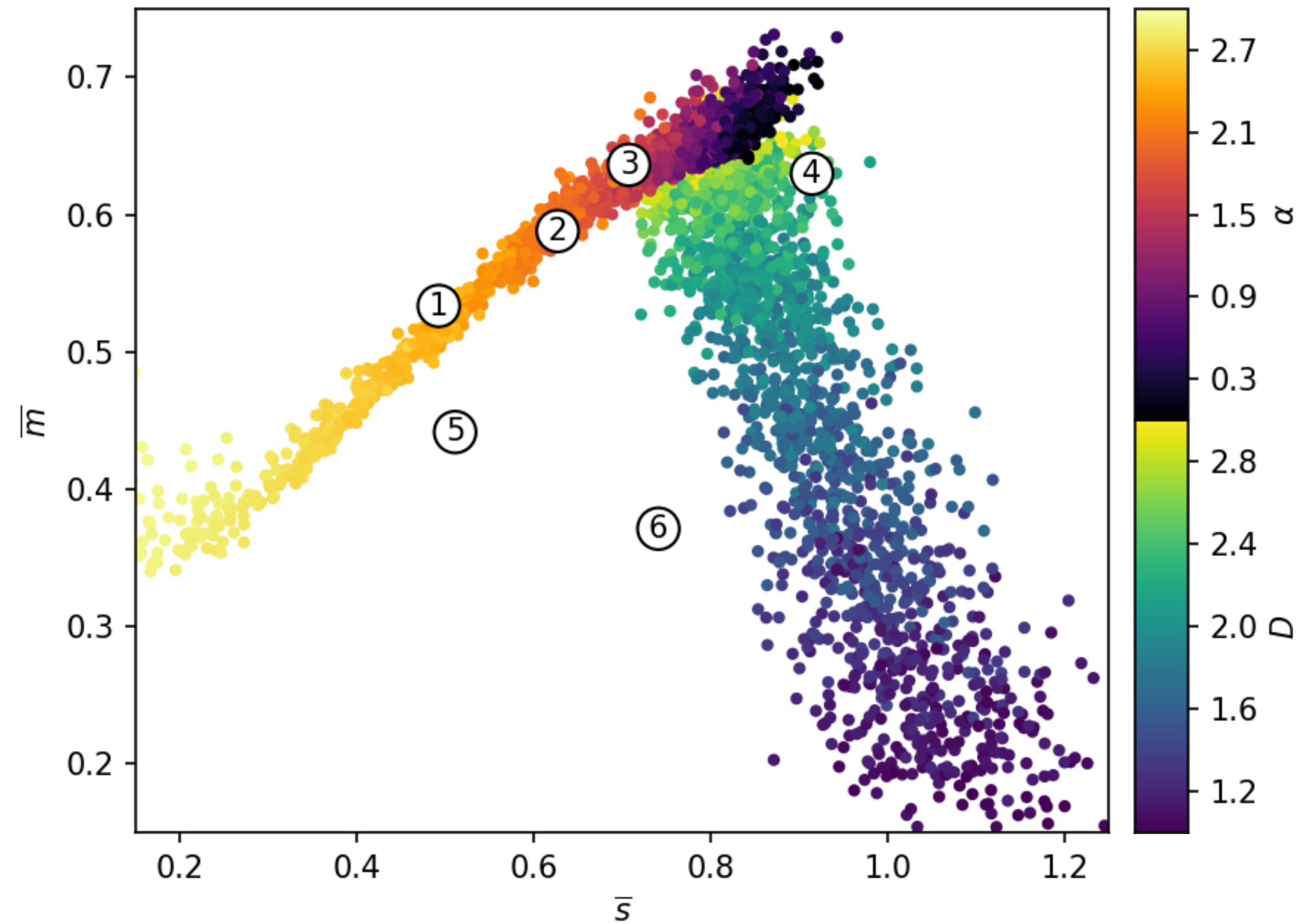


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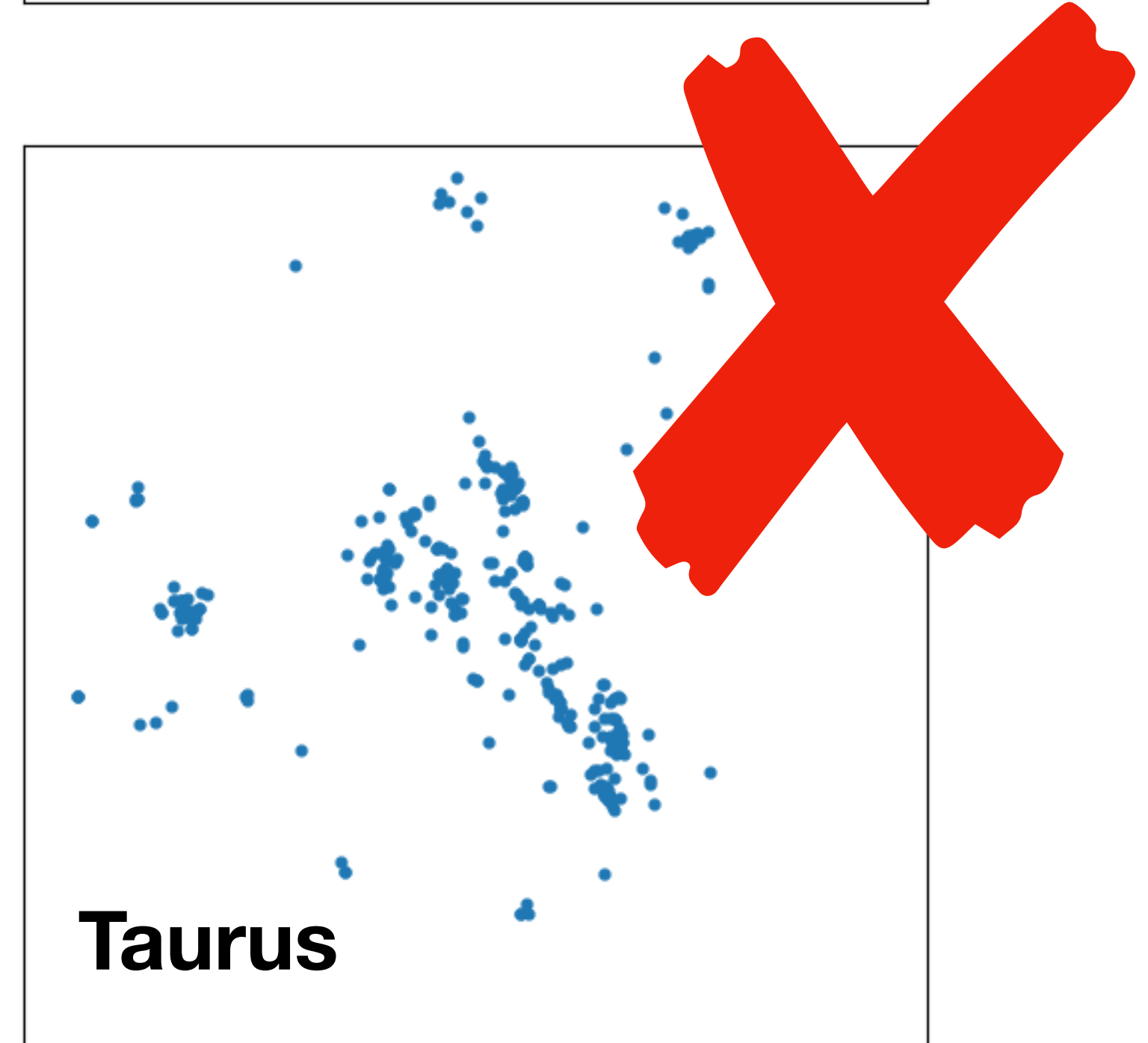
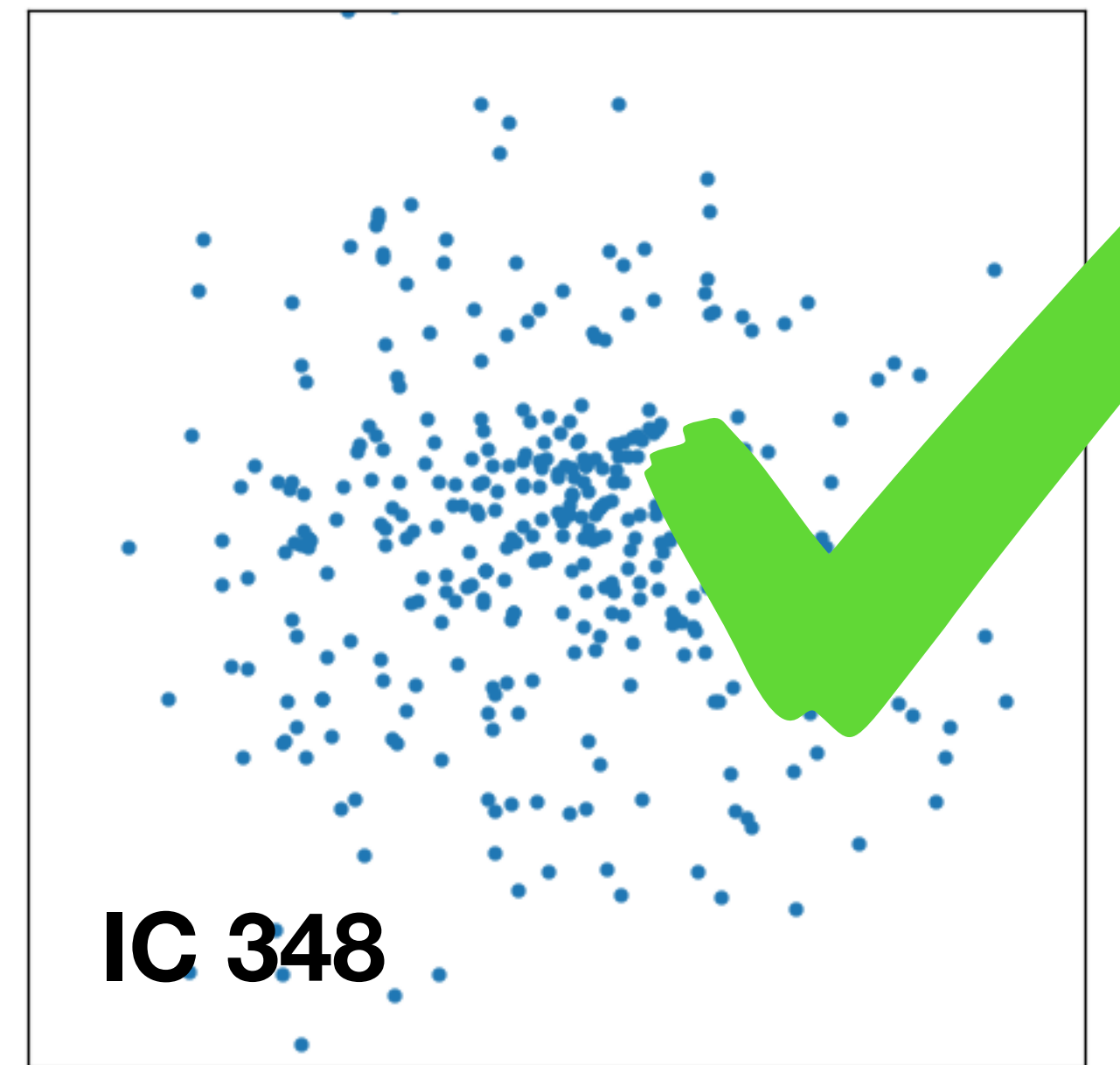
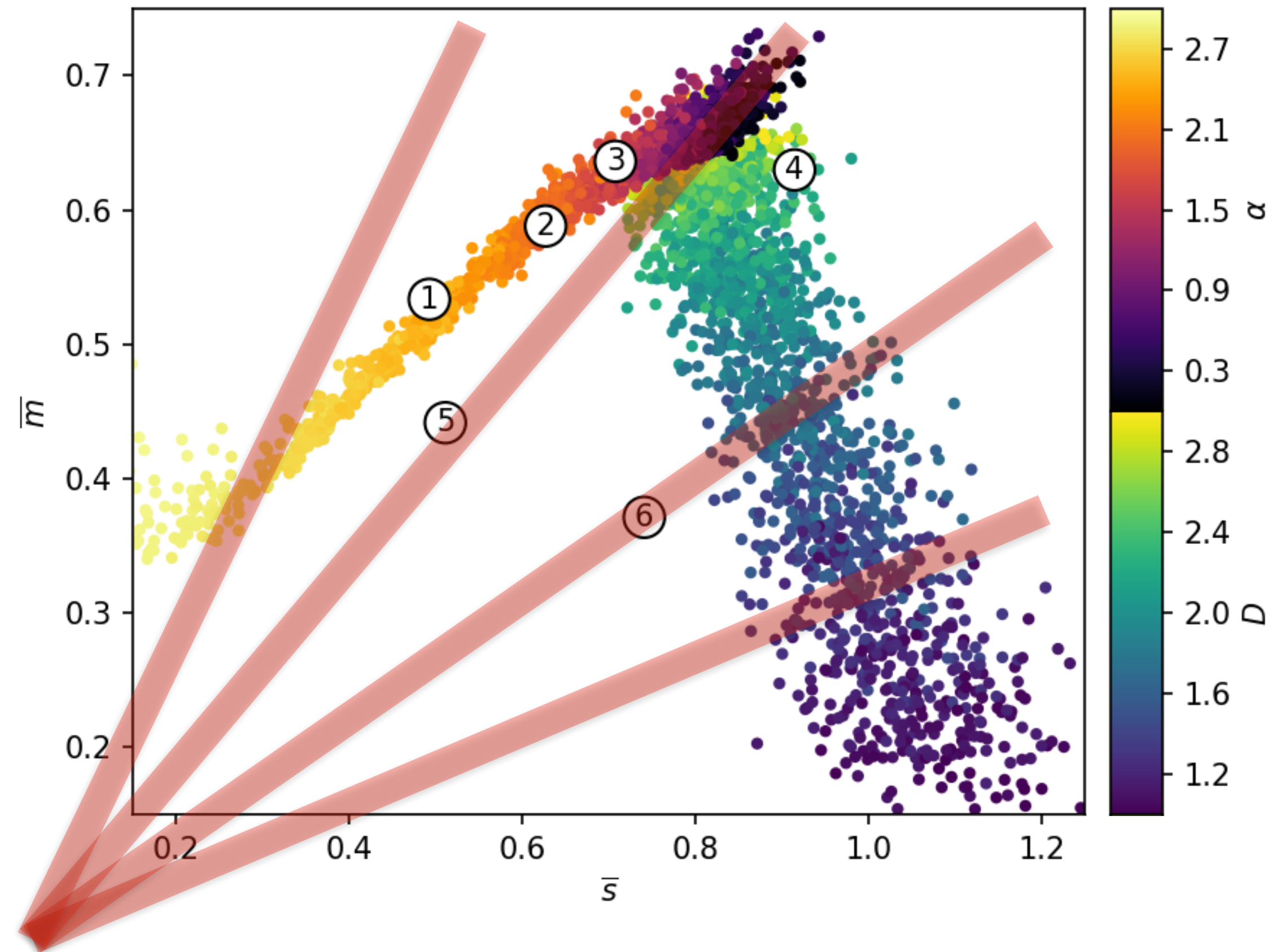


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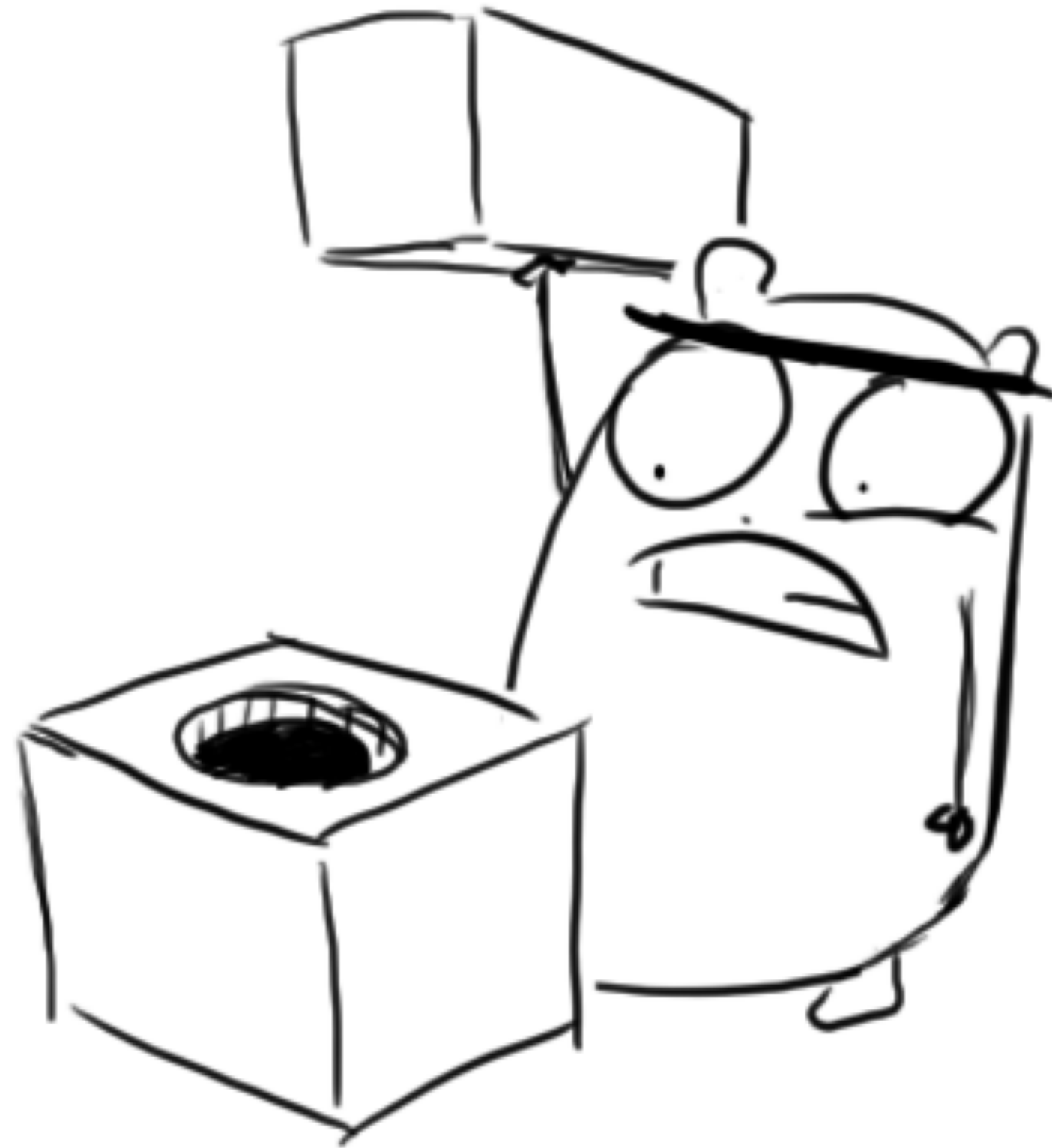
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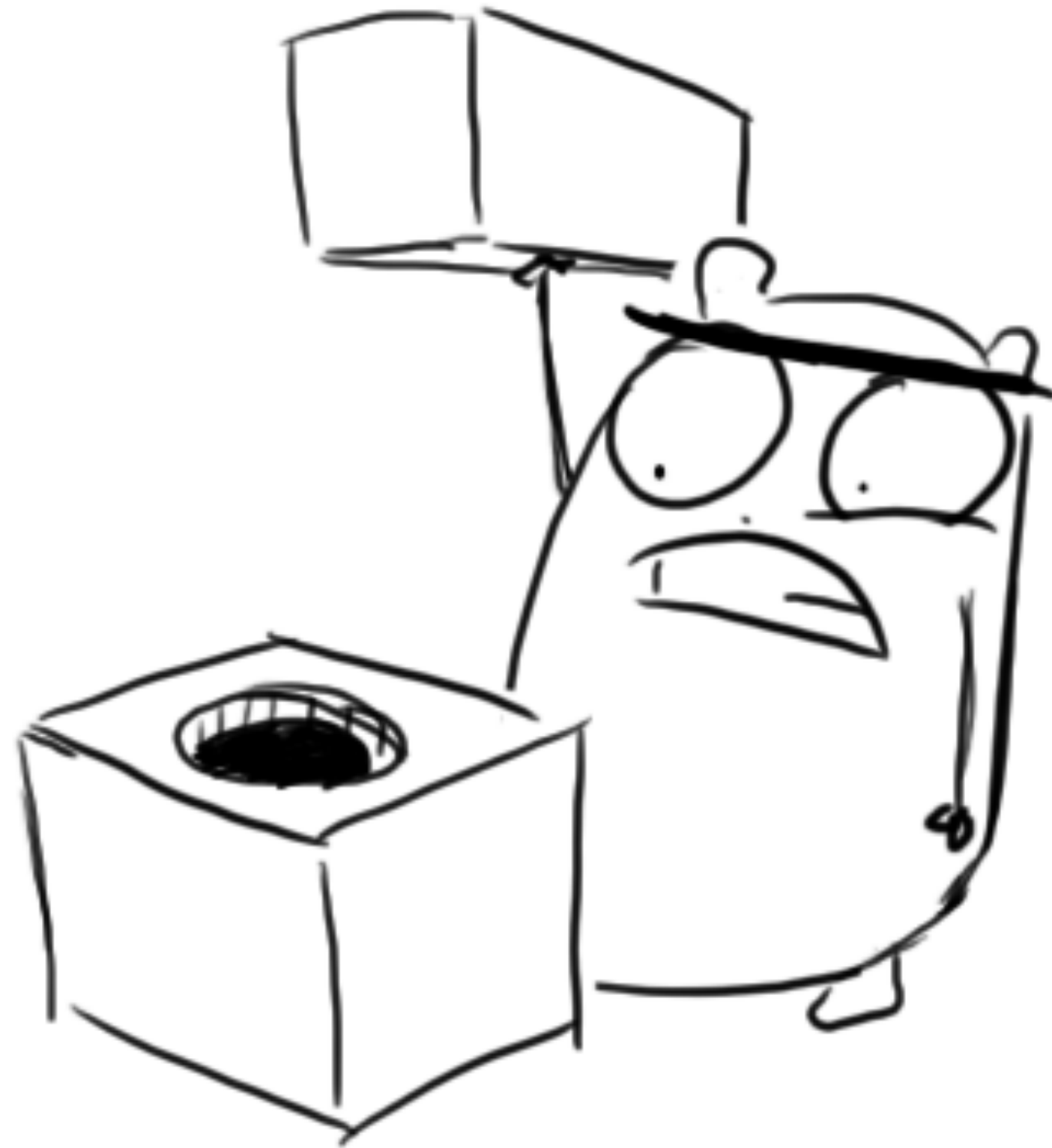
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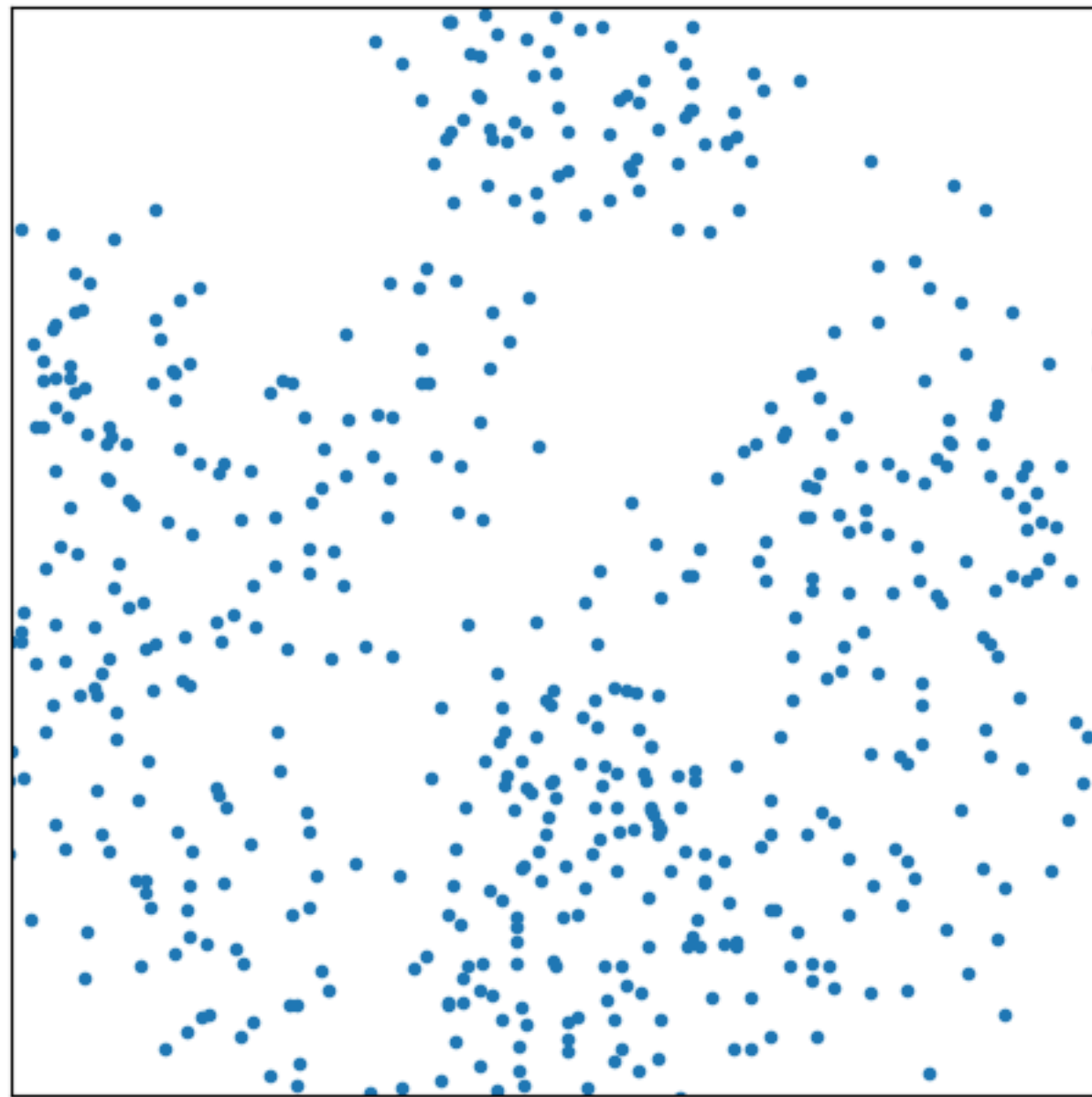
Stop using “square” models...



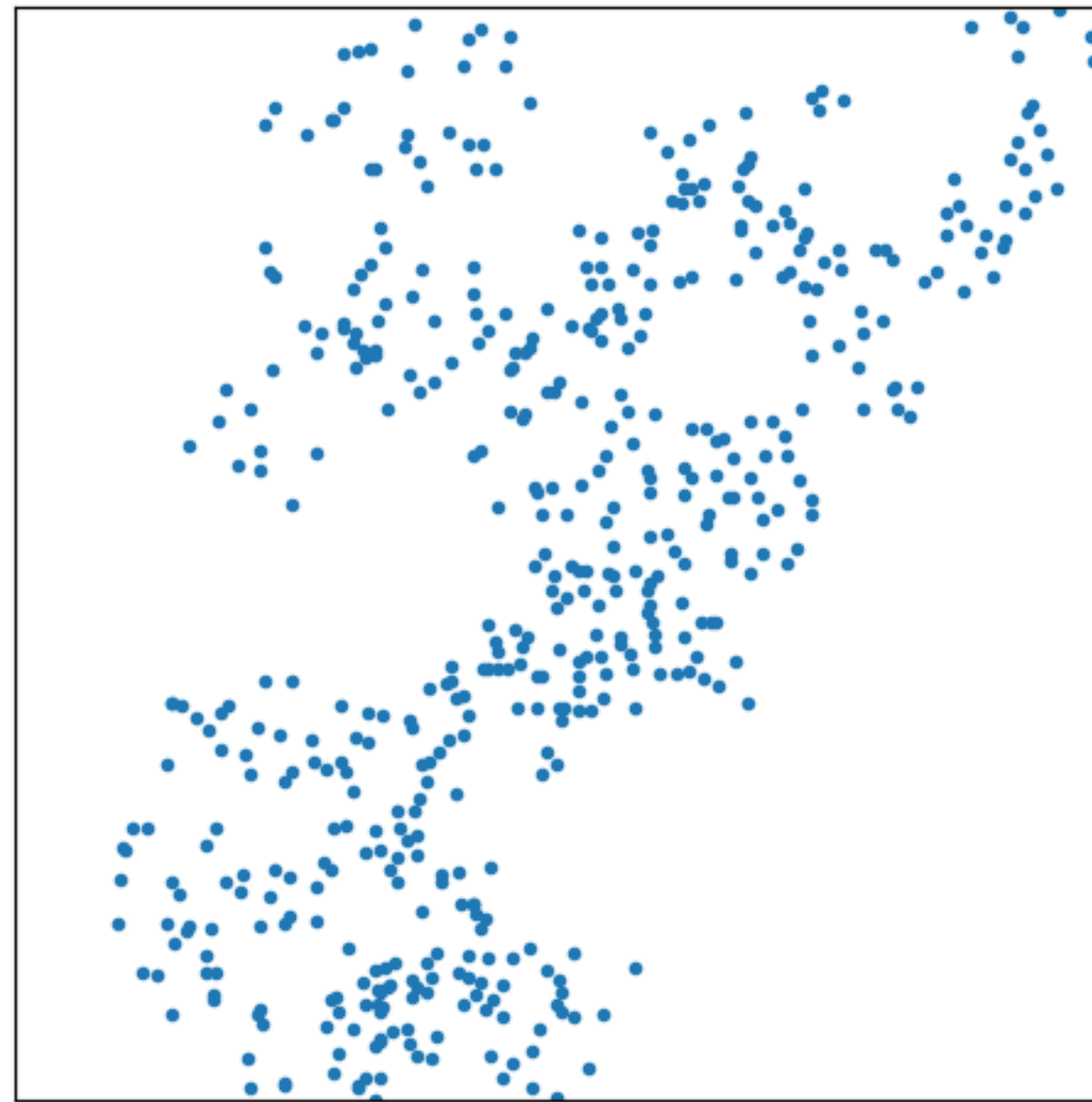
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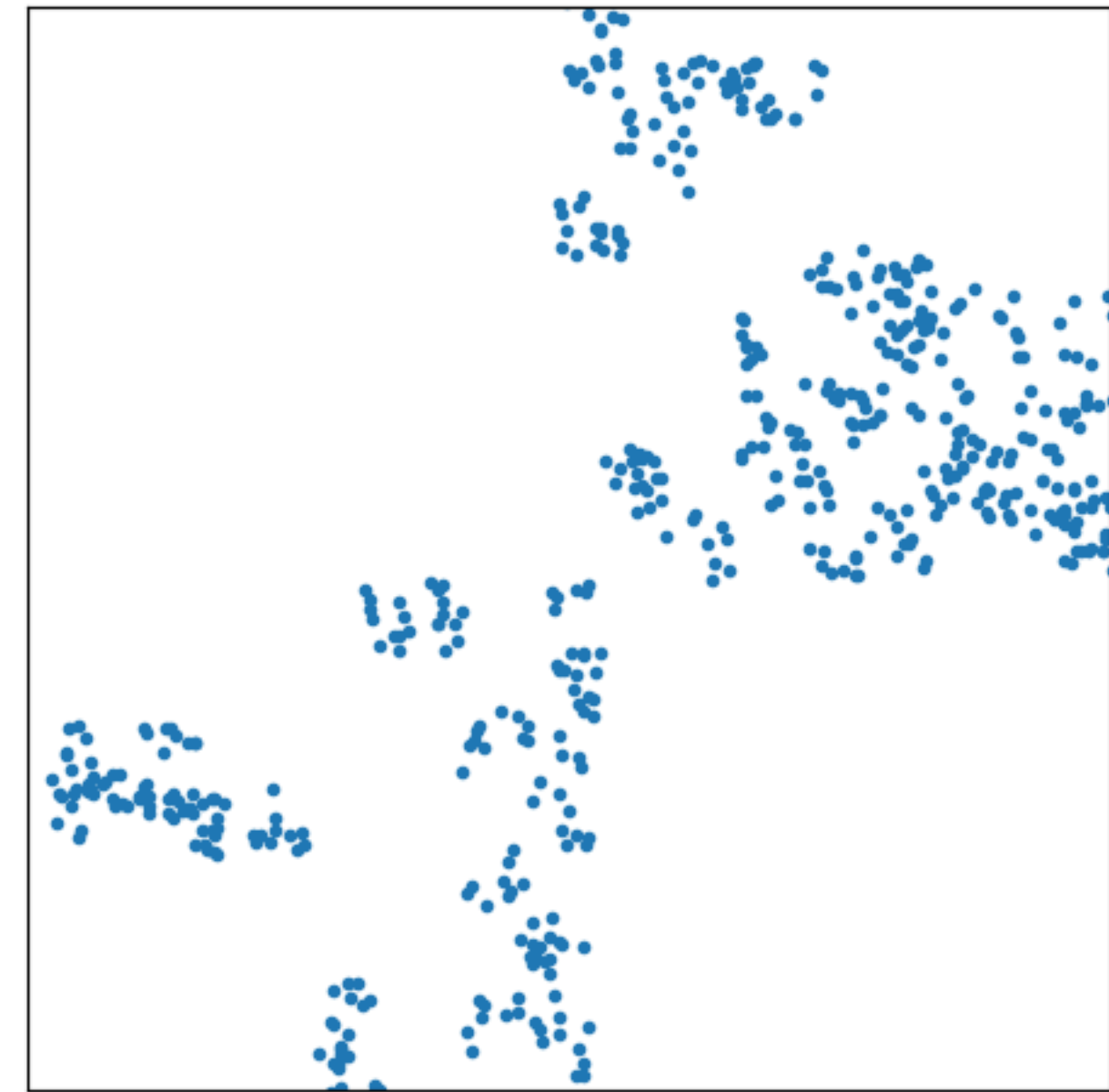
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$D=2.4$

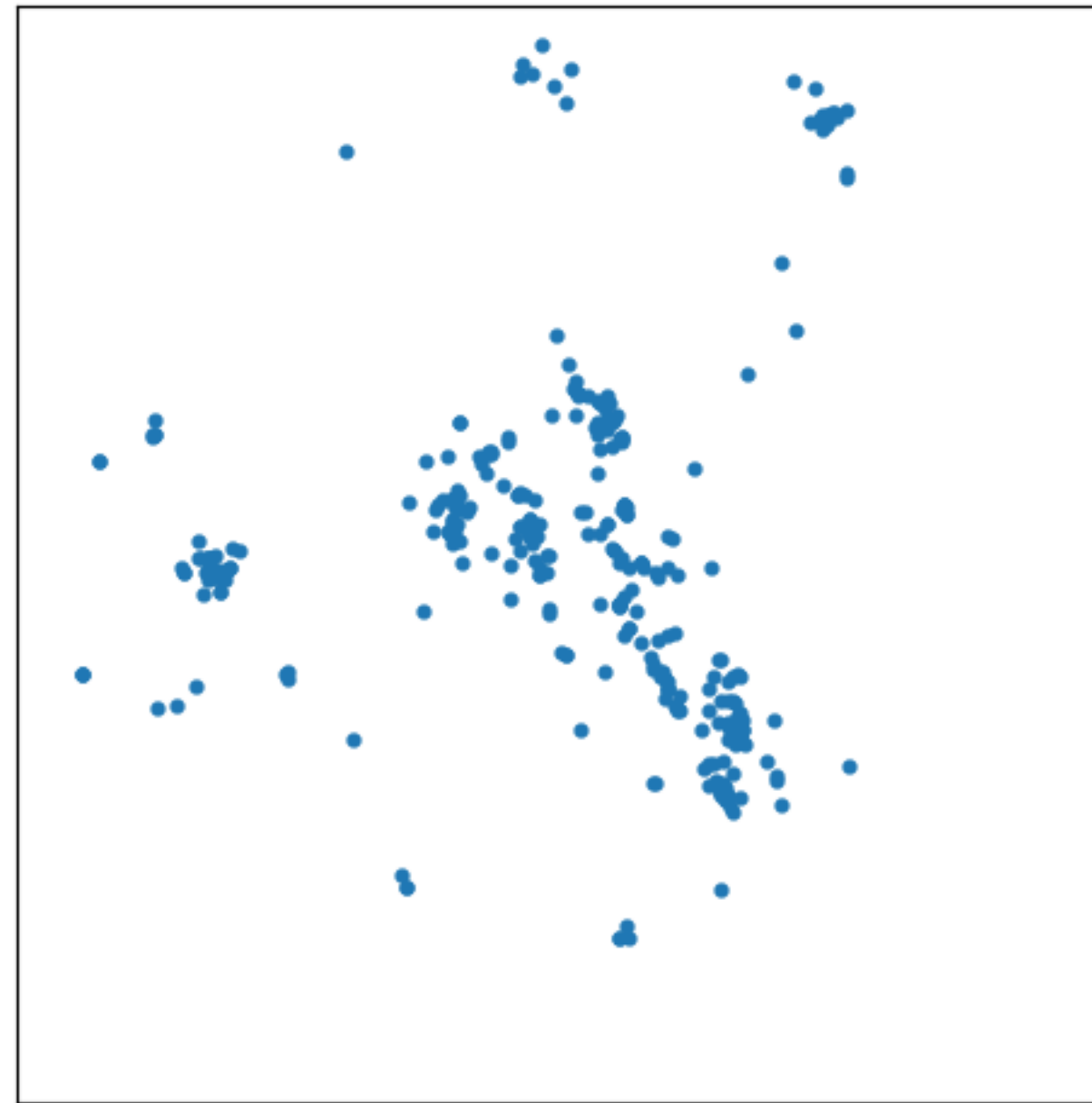


$D=1.8$

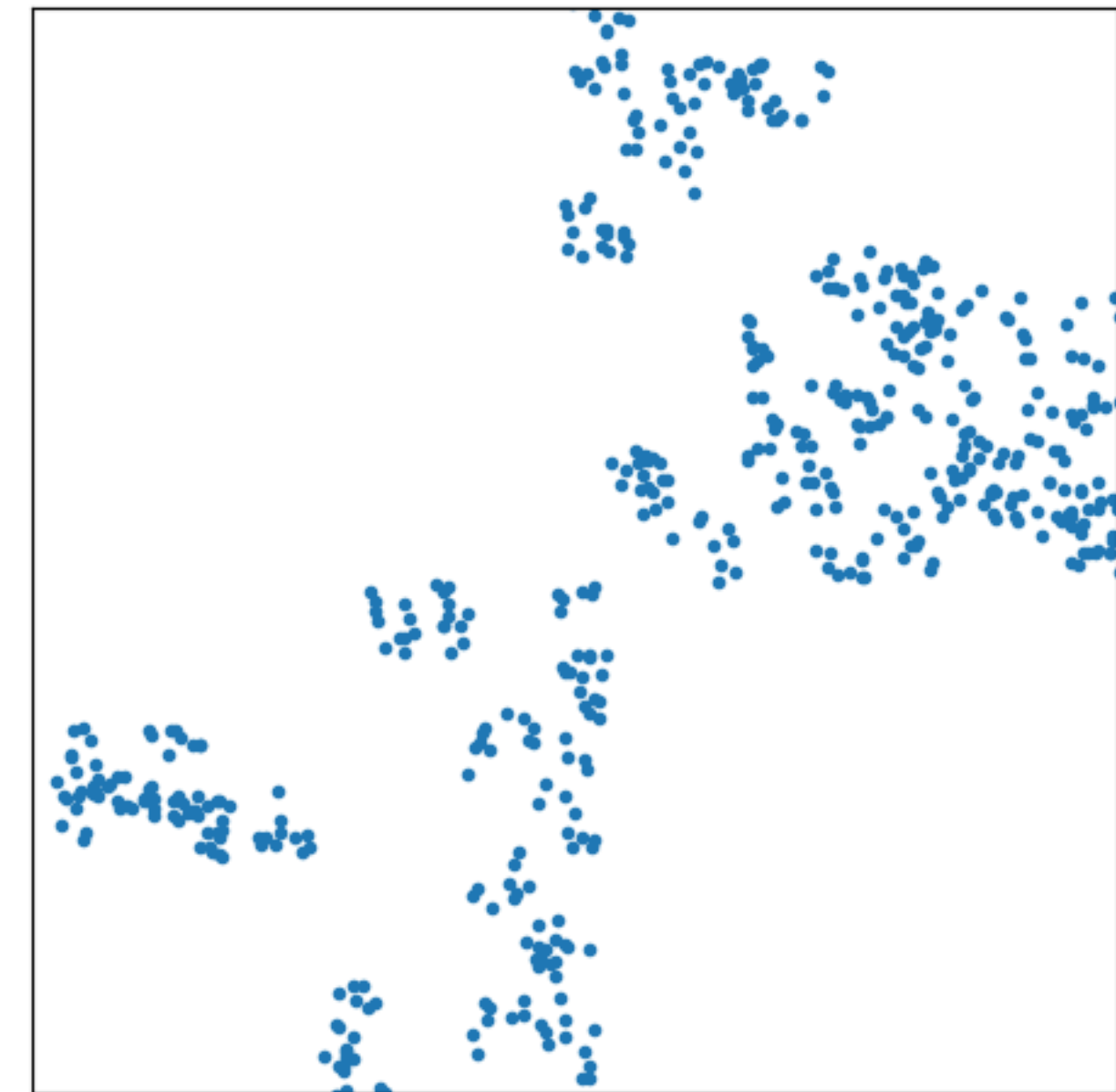


$D=1.5$

Stop using “square” models...



Taurus



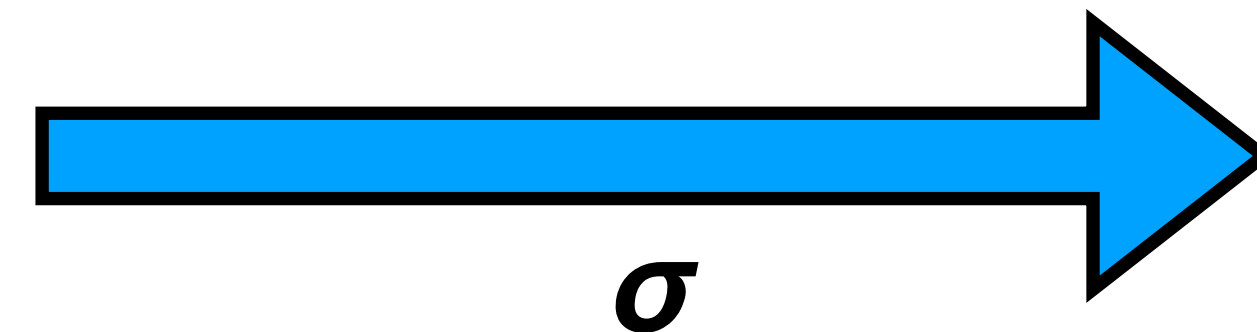
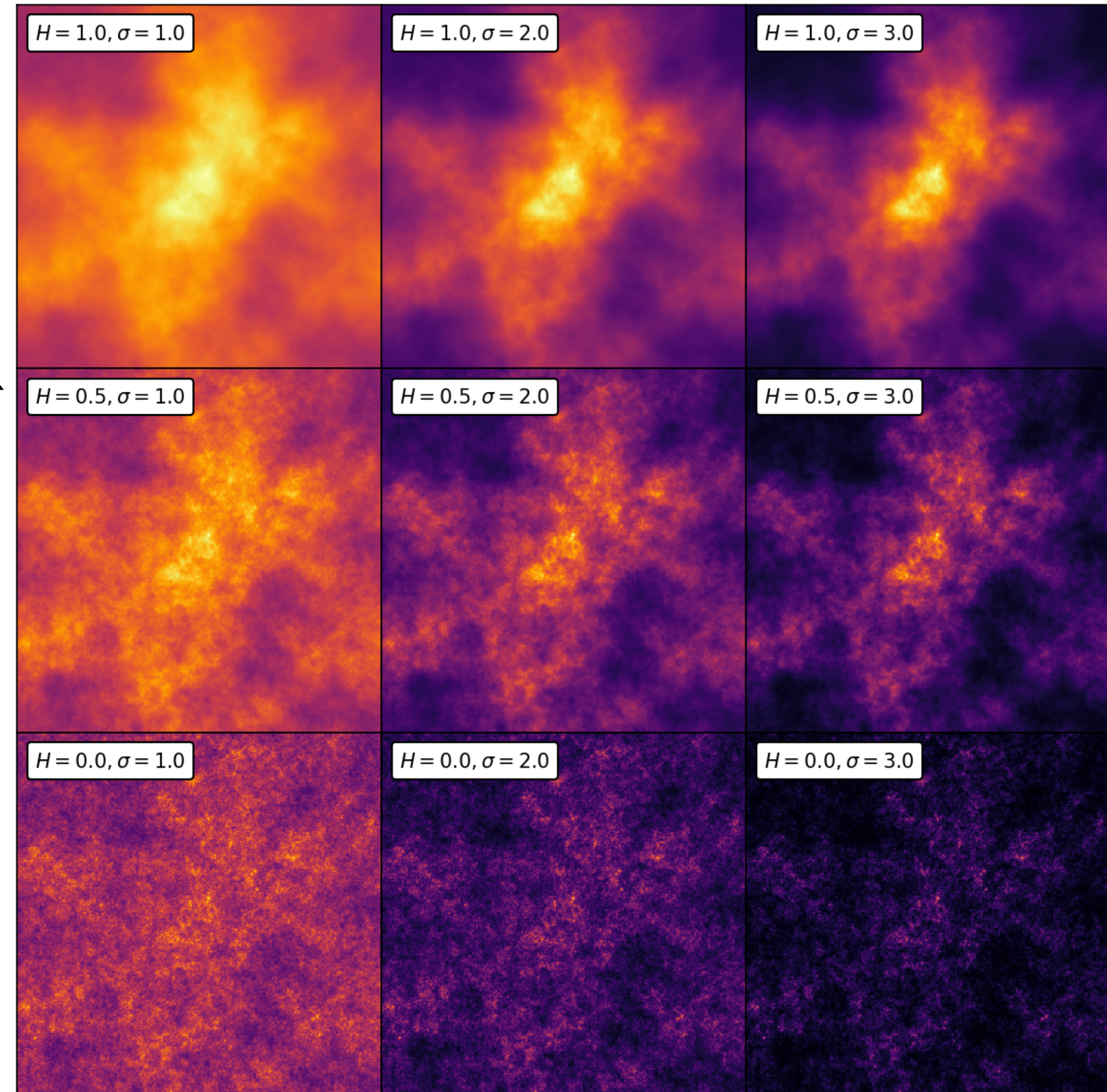
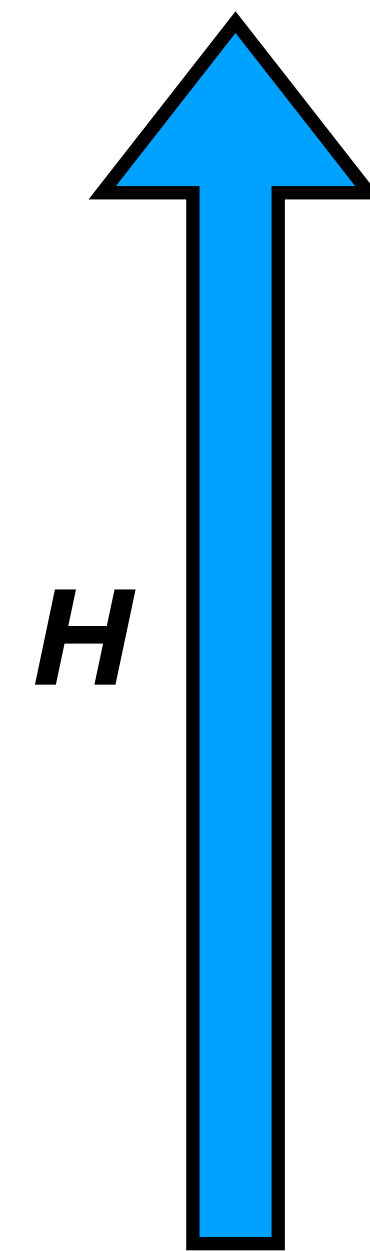
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Fractional Brownian Motion

- A generalisation of classical Brownian motion.
- Used to interpret properties of molecular clouds (e.g. Stutzki et al., 1998, Elia et al., 2014).
- Parameterised by the “Hurst index” H and a standard deviation σ .
- Relatively simple to generate in Fourier space:

$$P(k) \propto k^{-\beta}$$

$$\beta = E + 2H$$

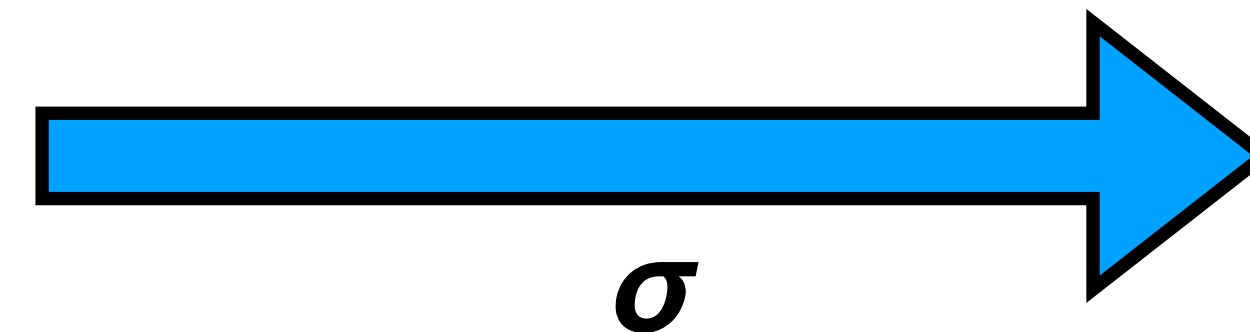
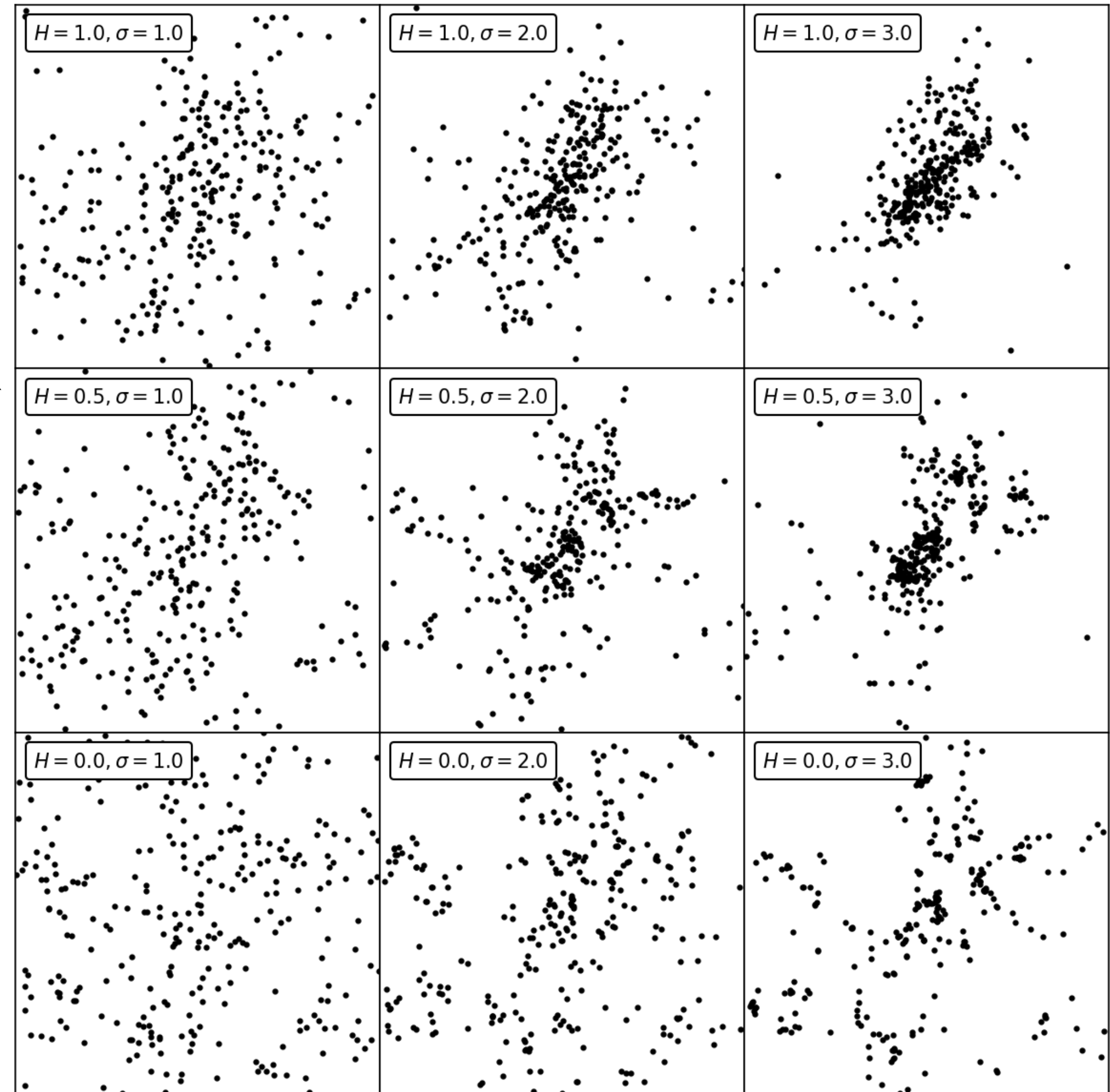
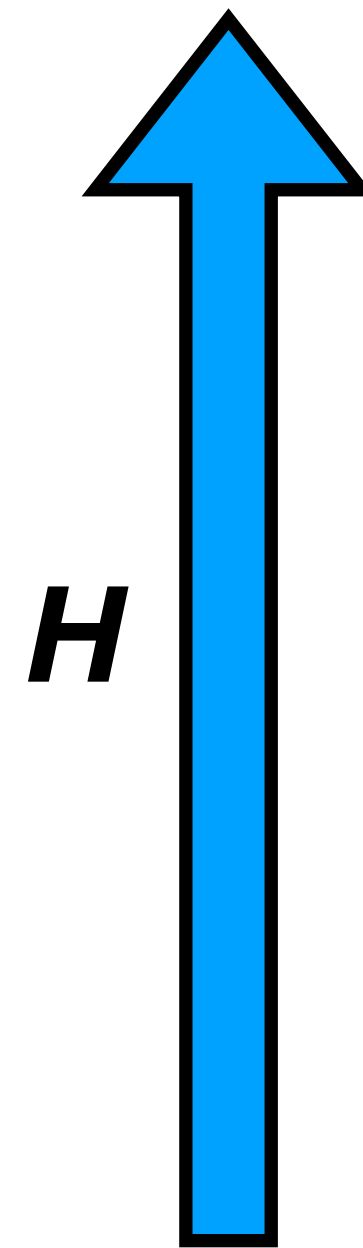


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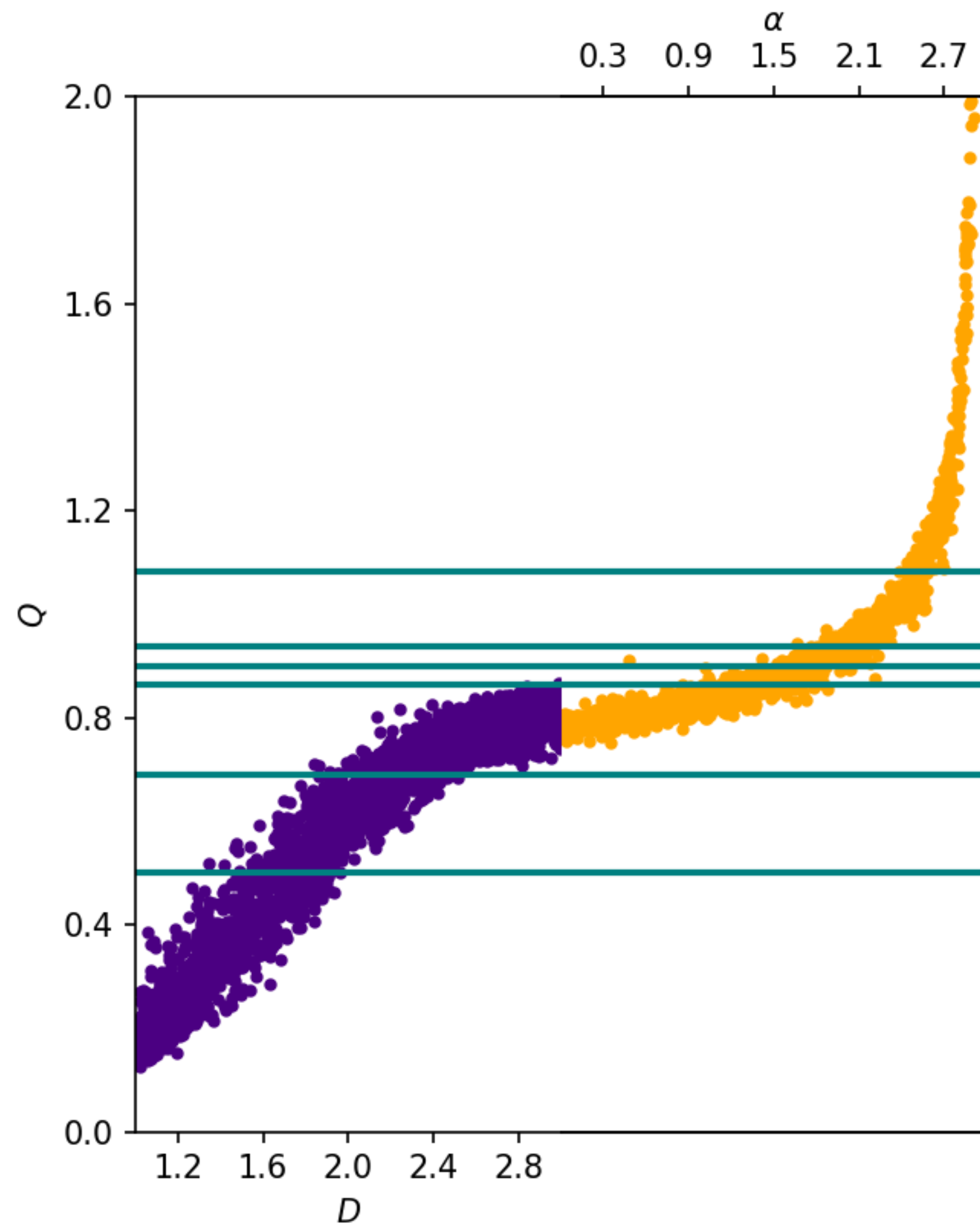
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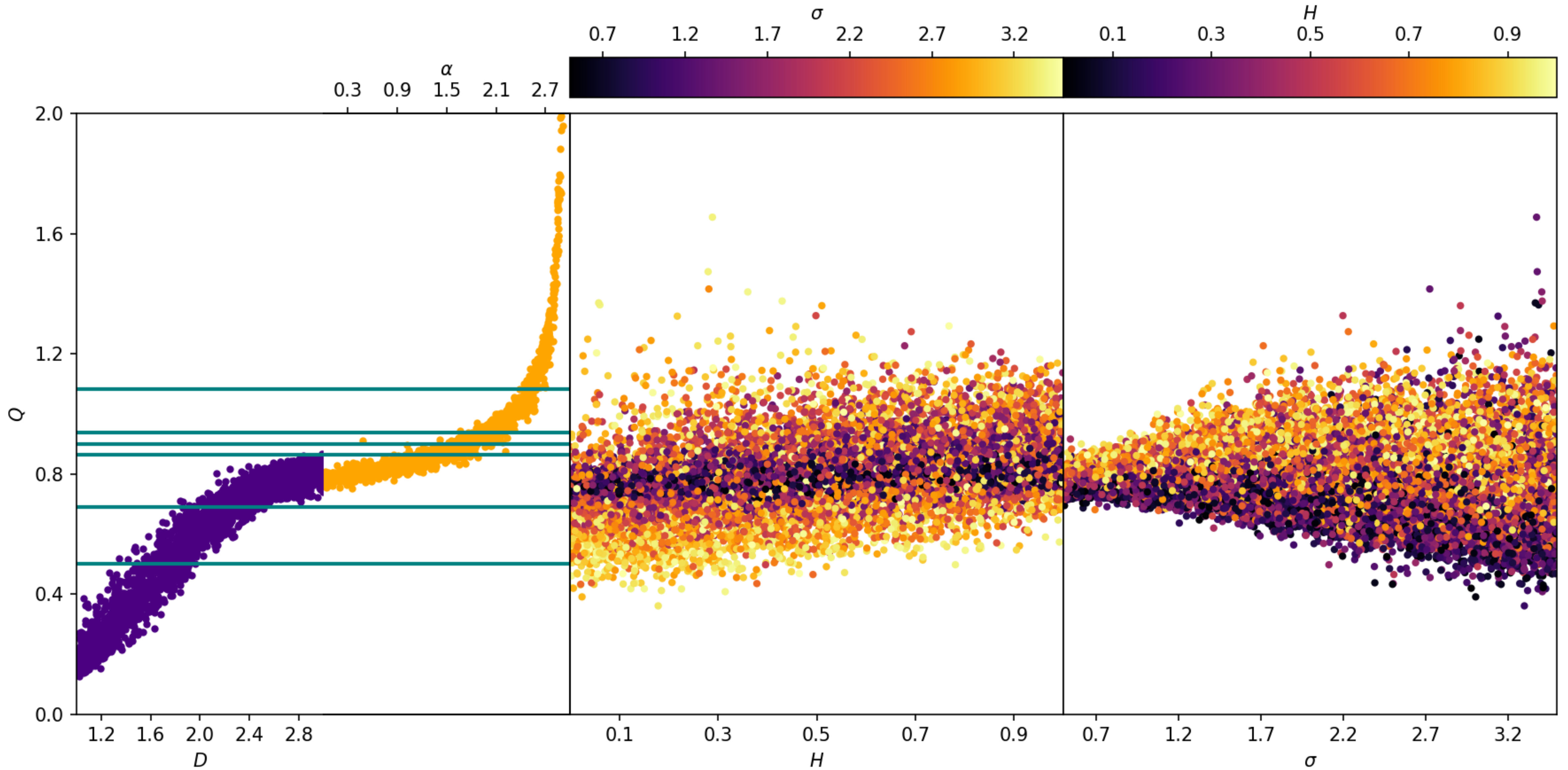
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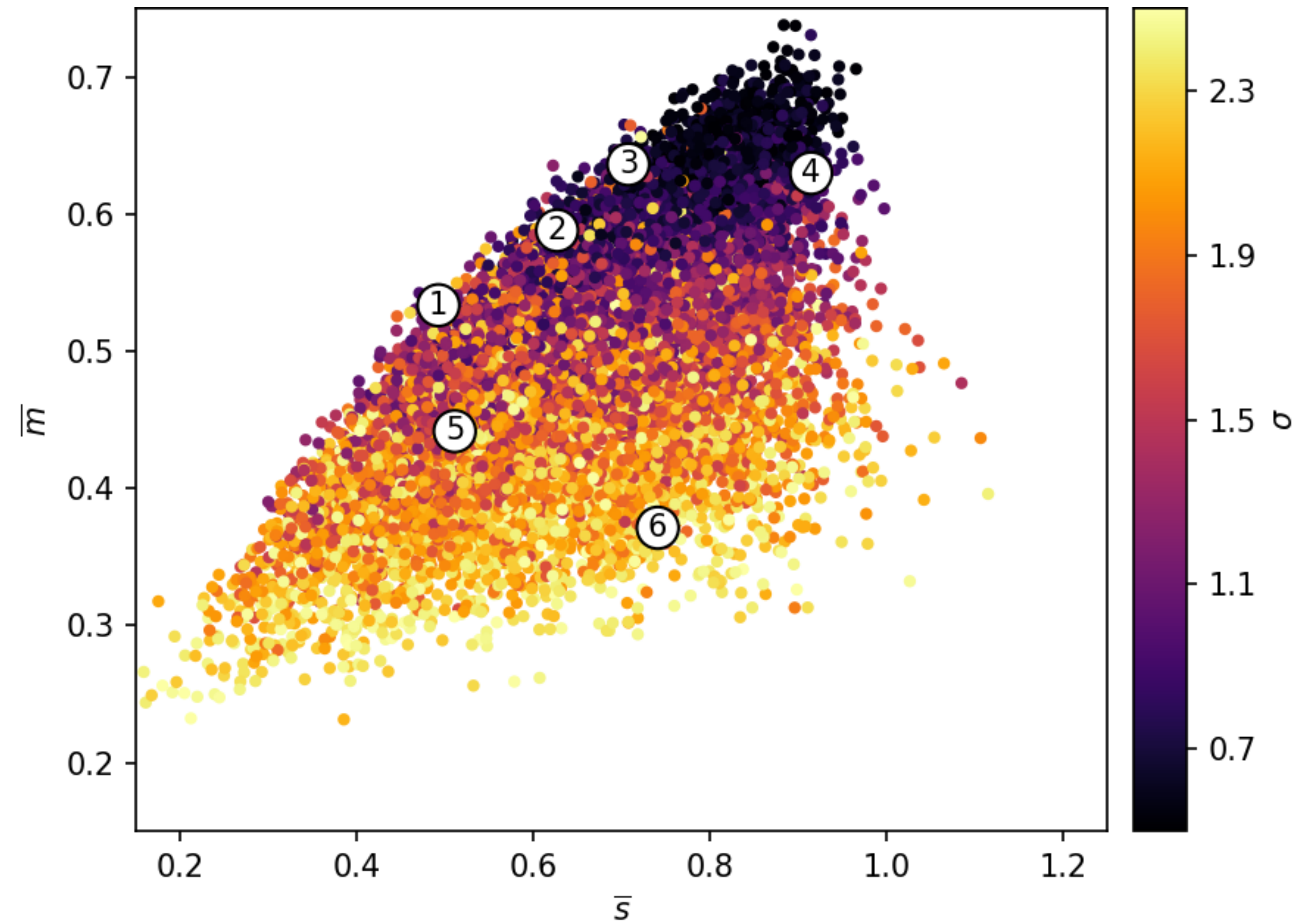
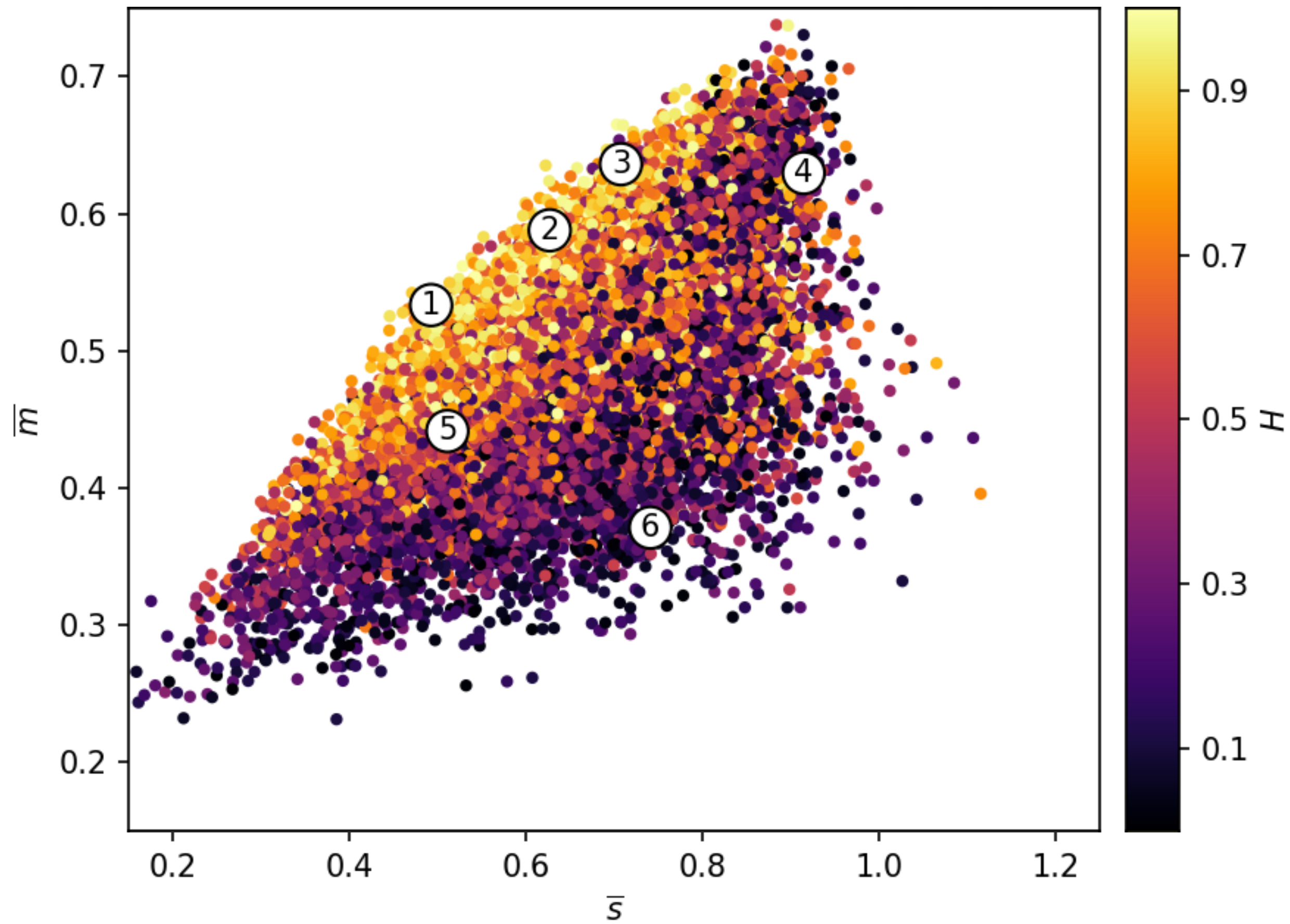
Q parameter



Q parameter is rubbish



\bar{m} - \bar{s} plots

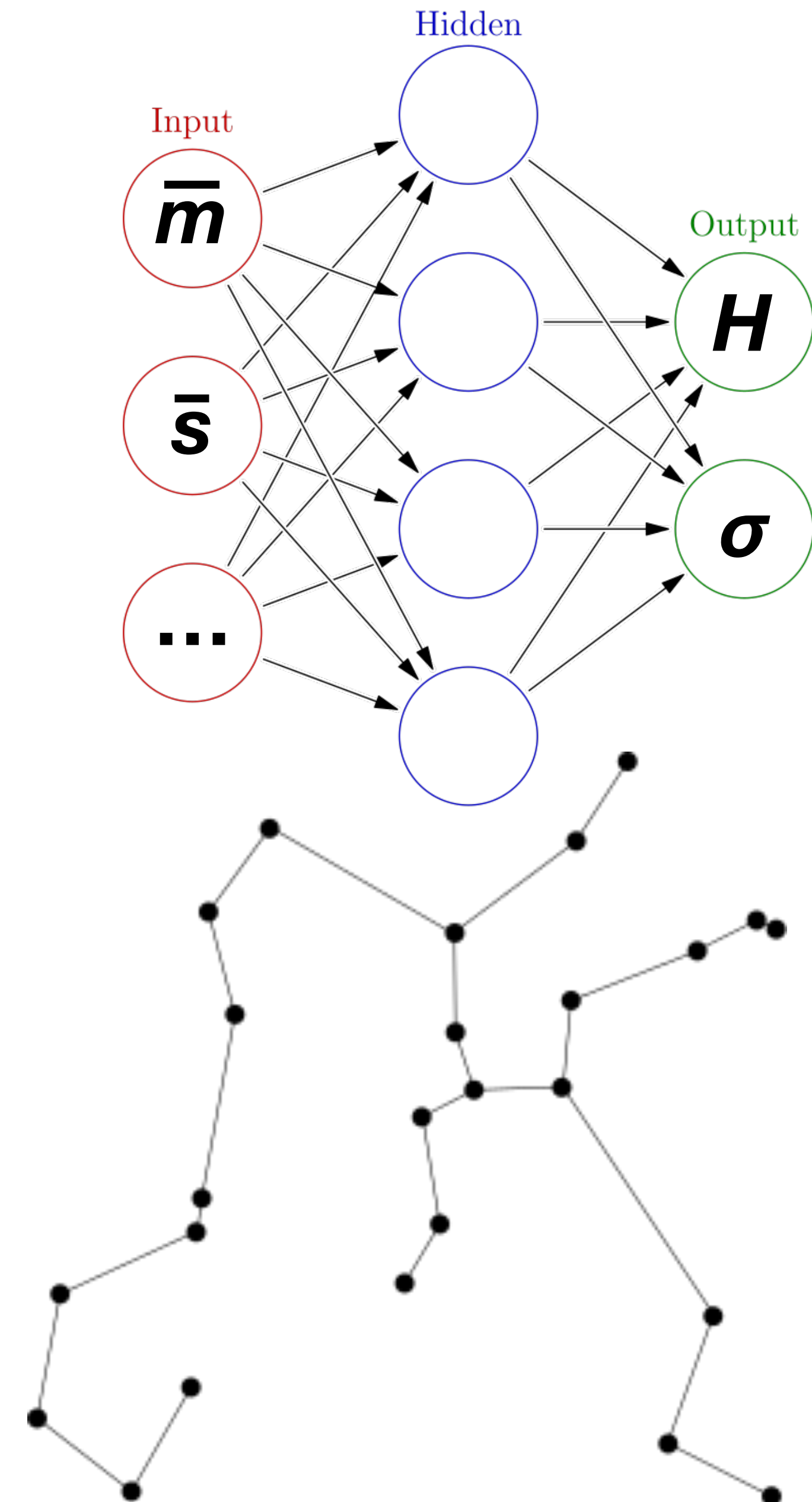


Machine learning

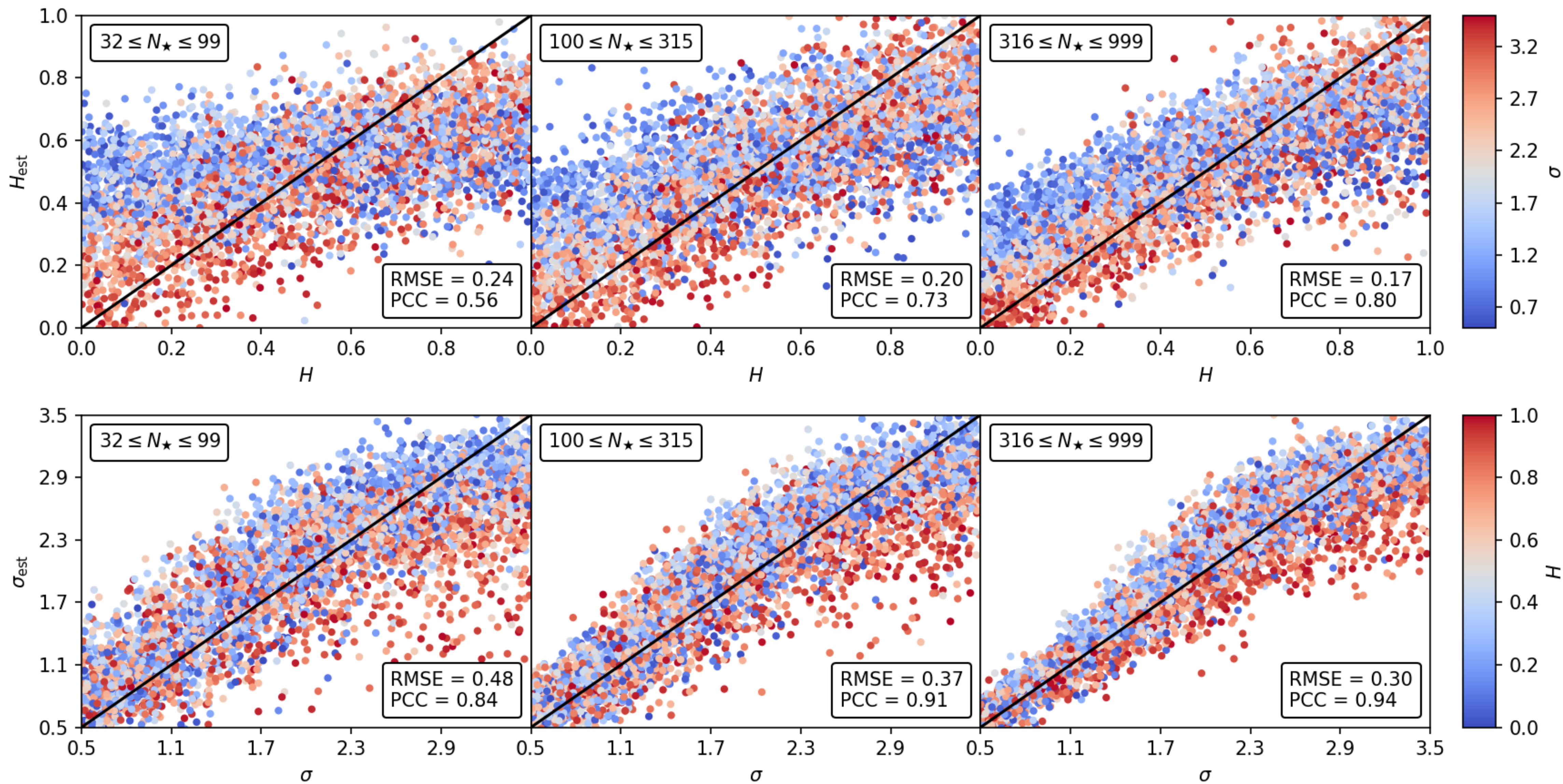
- Cluster distribution is “whitened”, i.e. $\text{Cov}(\mathbf{x})=\mathbf{I}$
- Parameters estimated using minimum spanning tree and complete graph.
- Neural net method used to map edge length moments of graphs to underlying parameters.

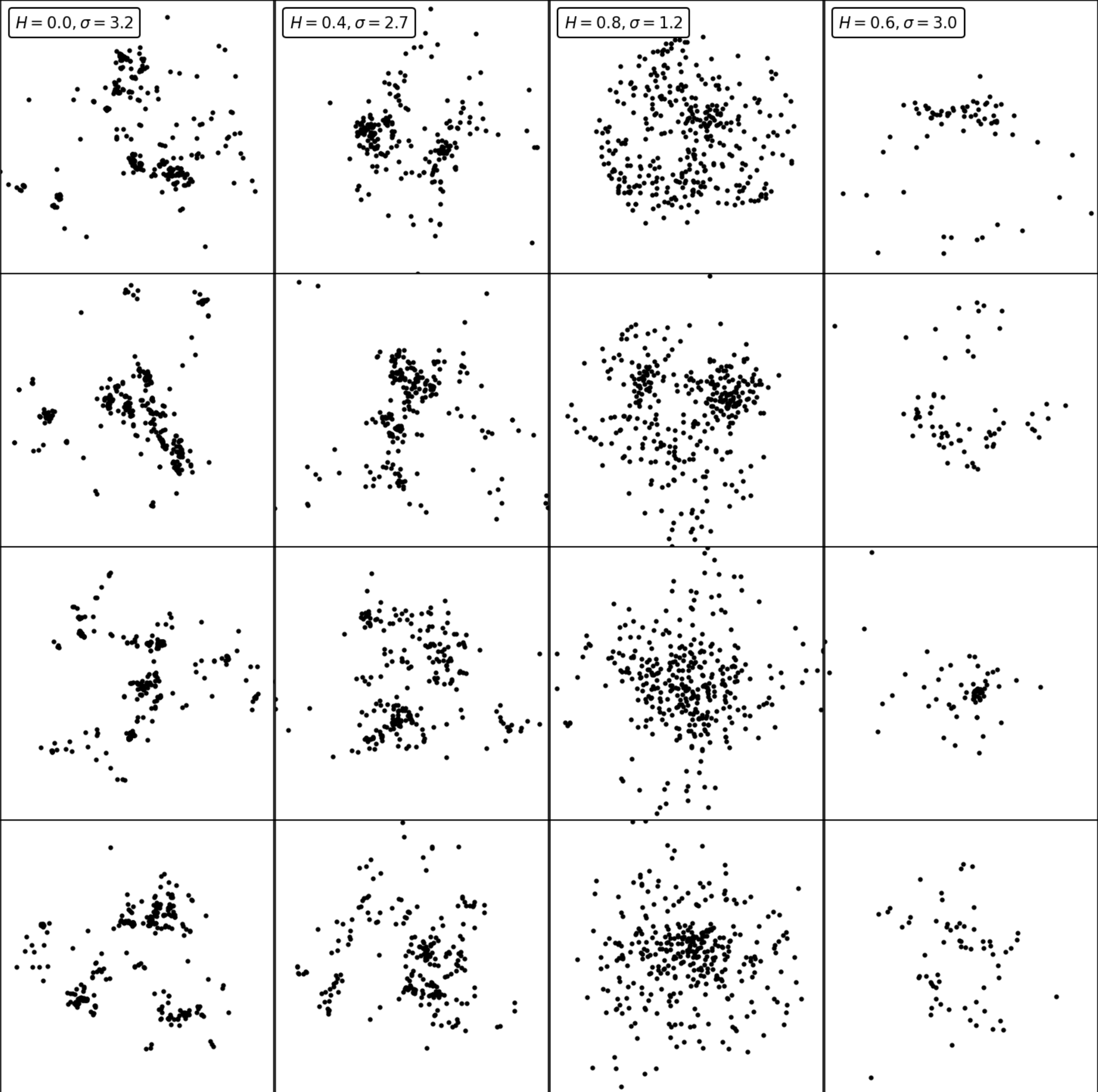
$$M_n(\mathbf{x}) = \left(\frac{1}{N} \sum_{i=1}^N (x_i - \mu_{\mathbf{x}})^n \right)^{1/n}$$

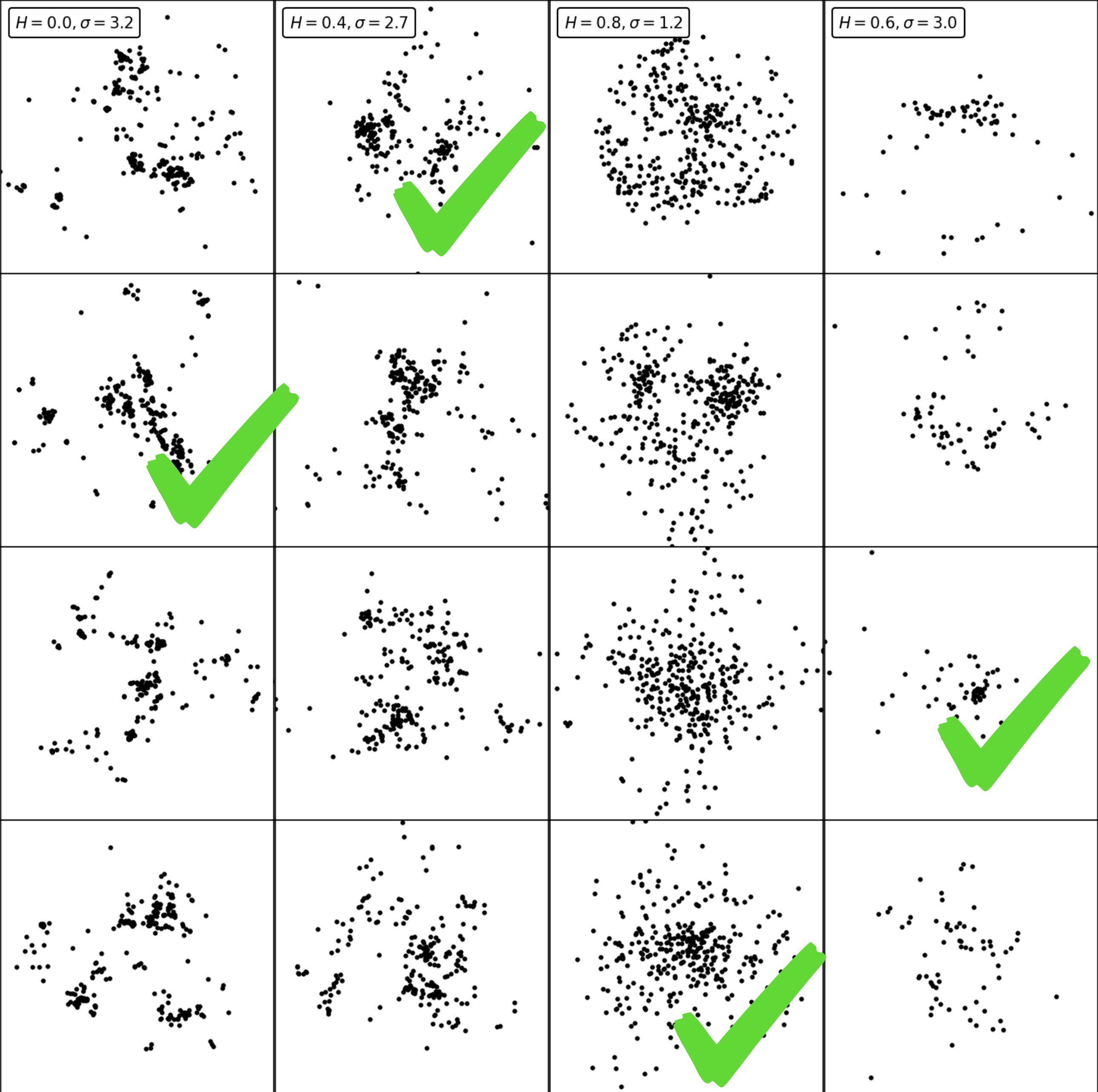
GitHub: github.com/odlomax/clusterfrac



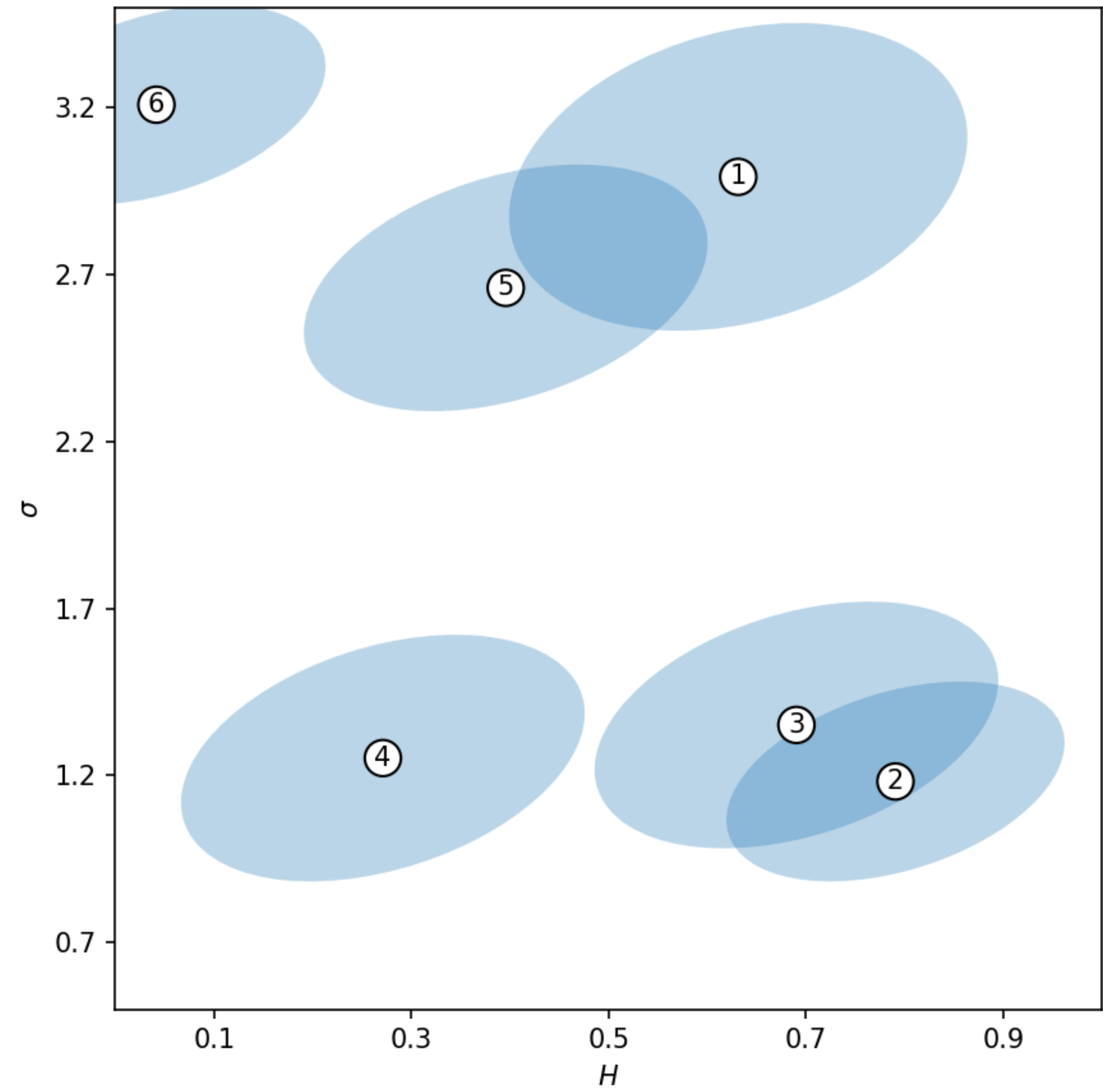
Parameter Estimation





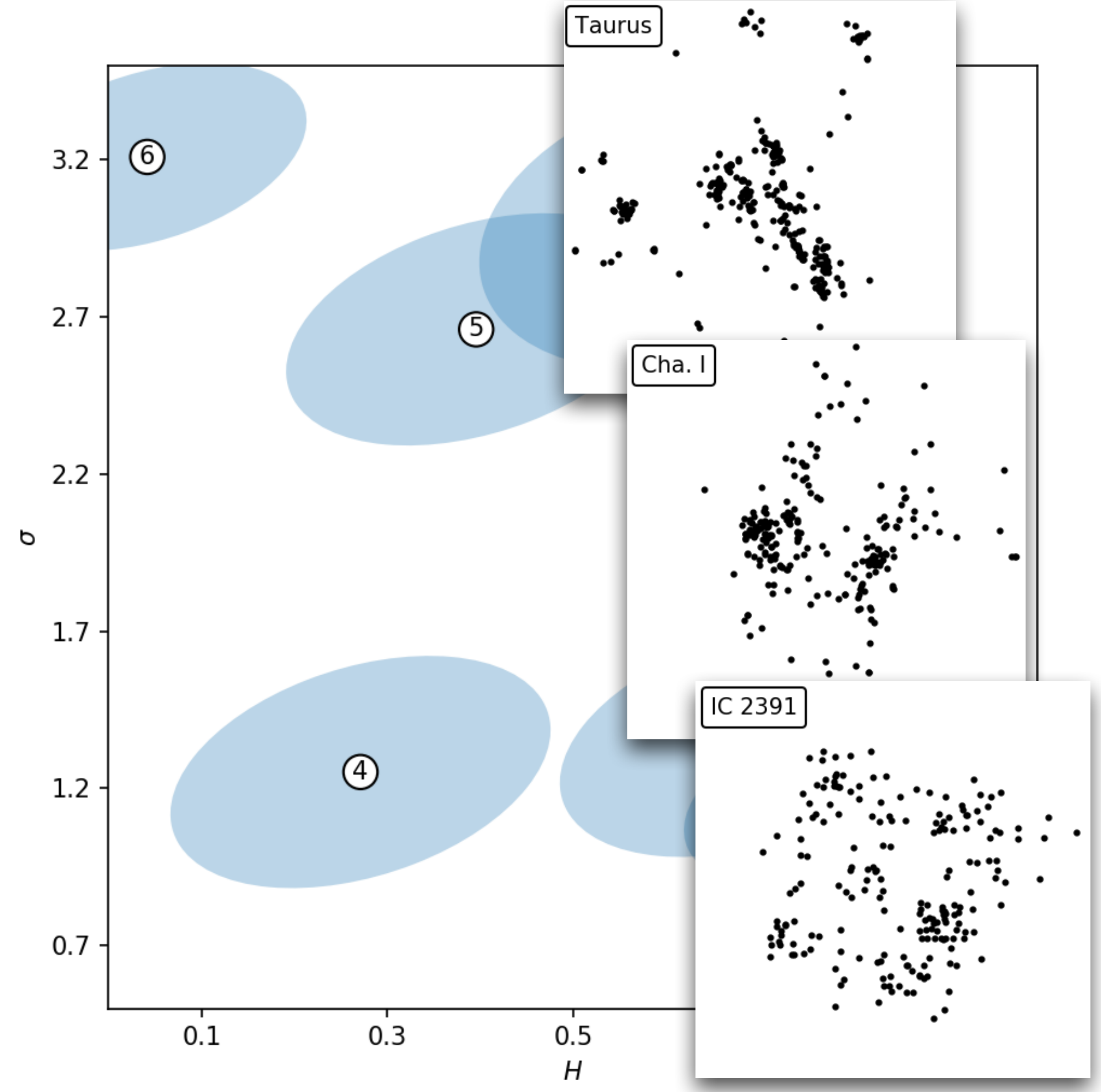


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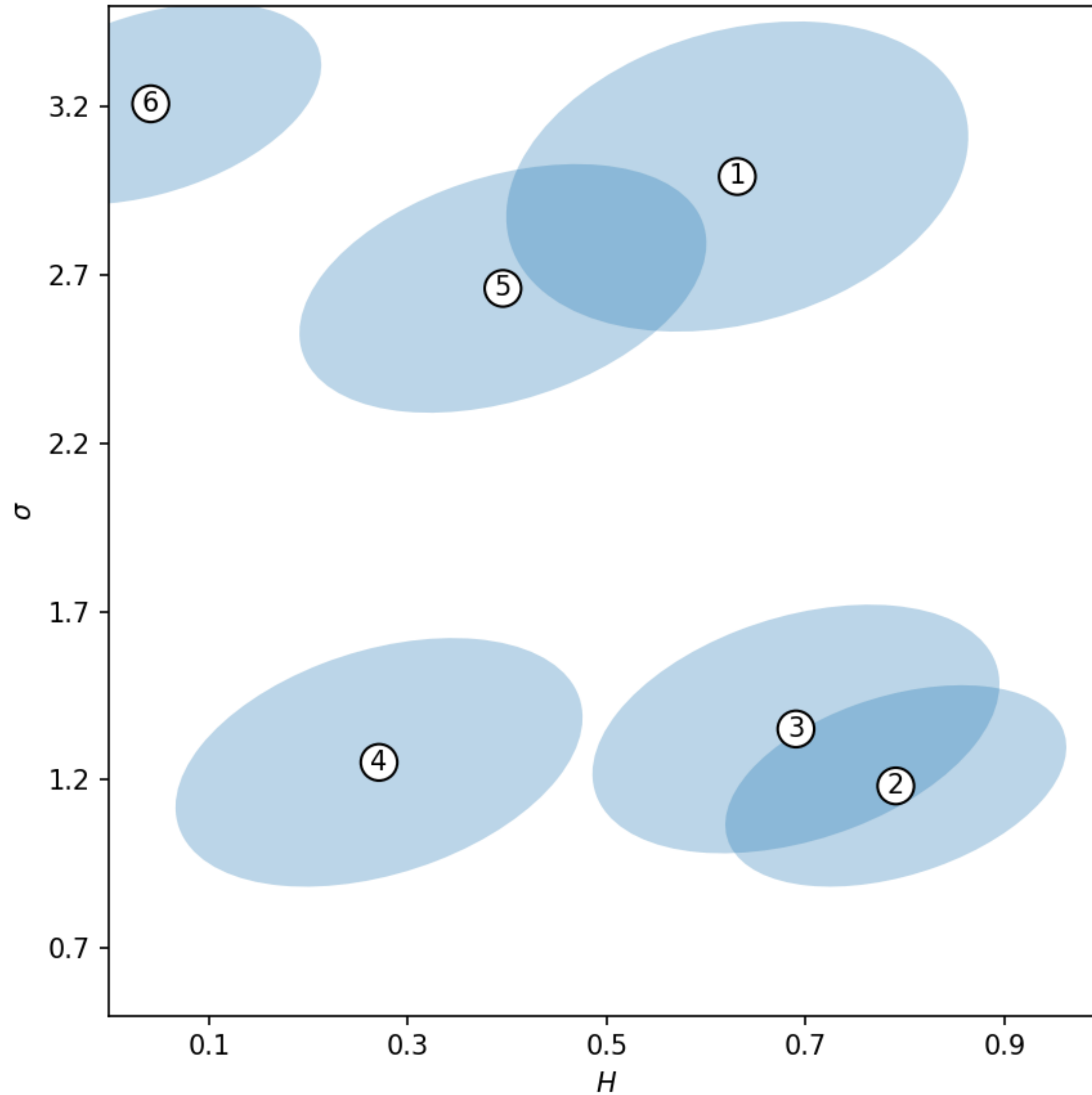
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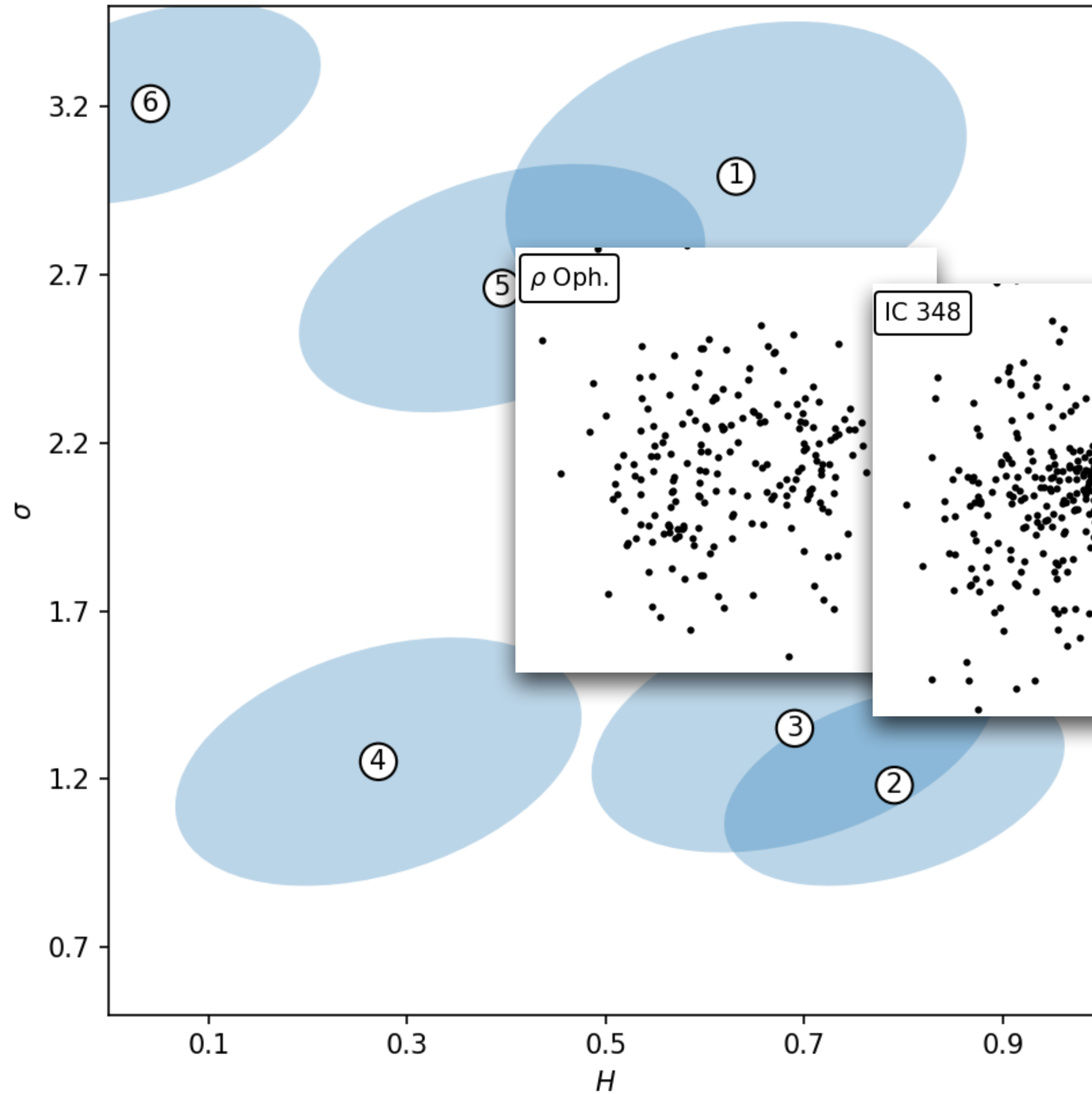
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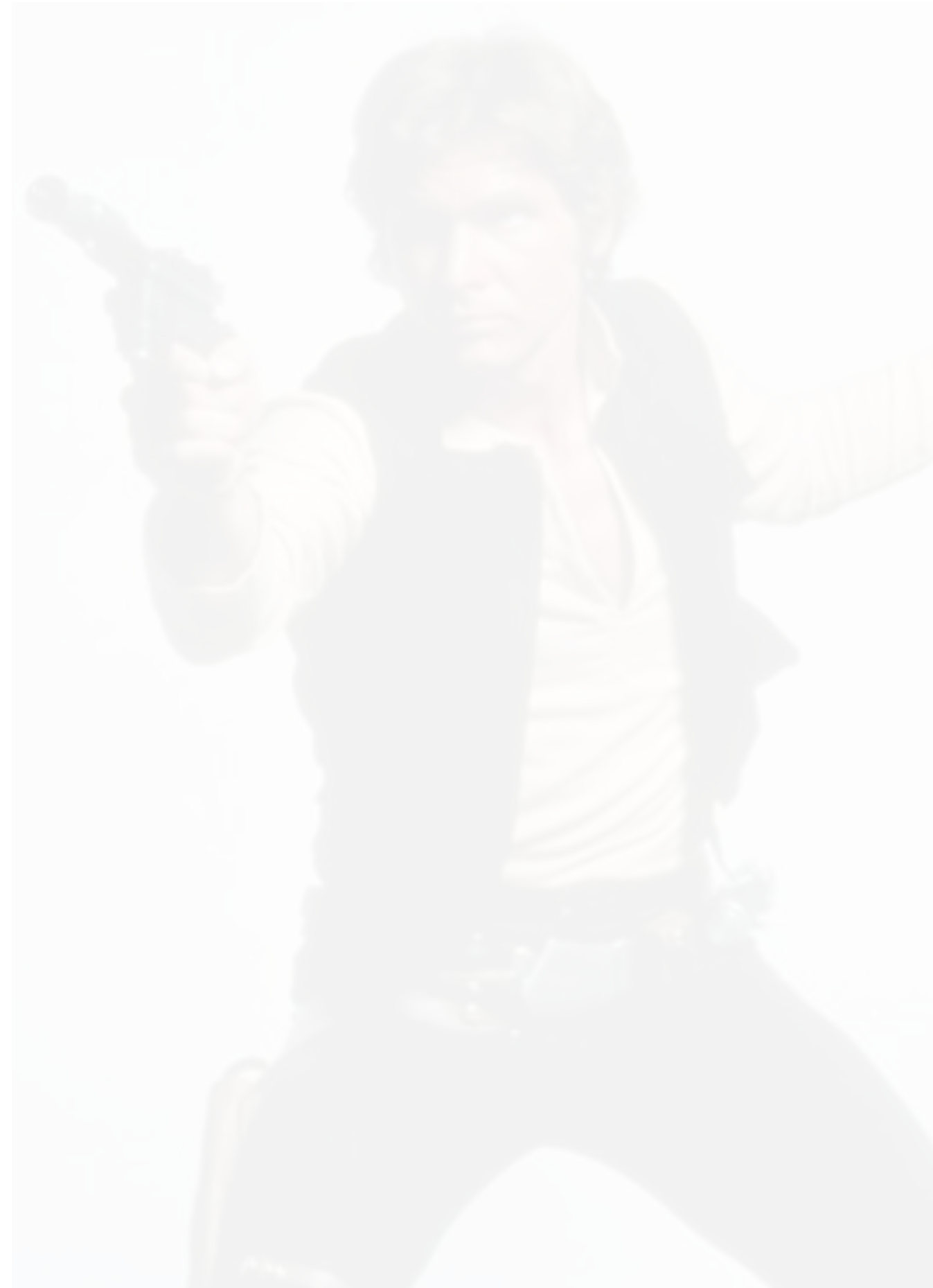
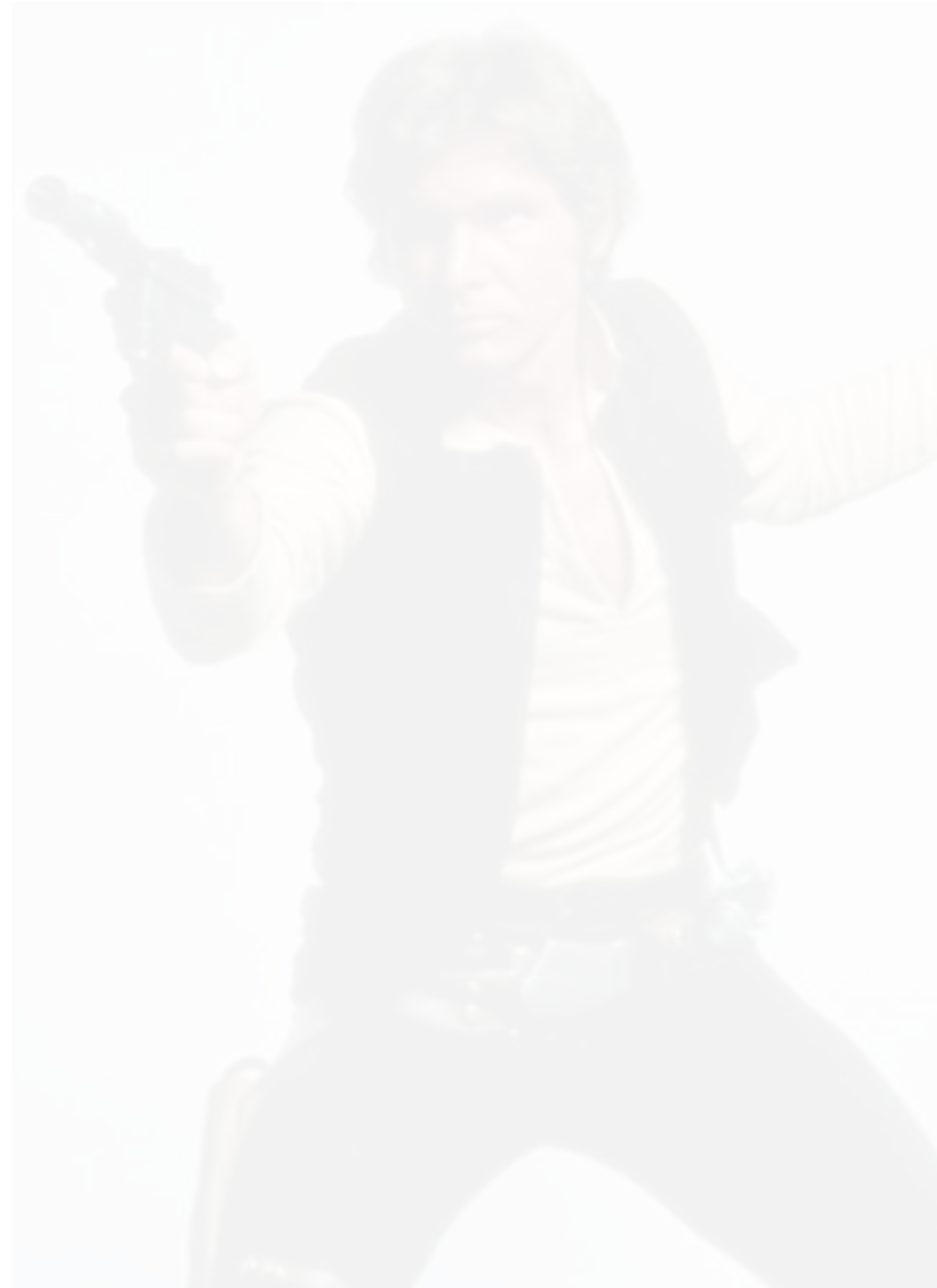
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HANS!



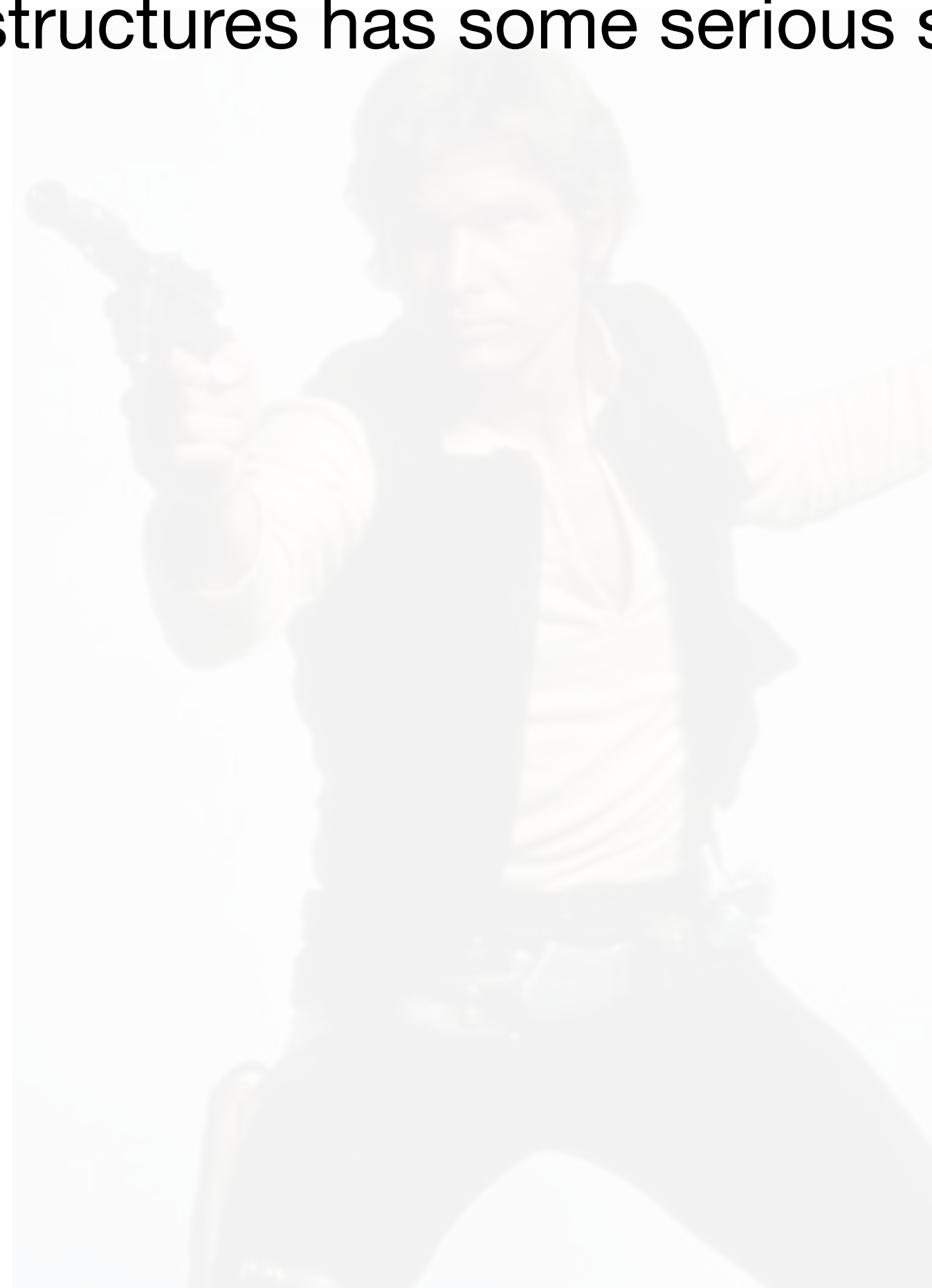
HANS!

Conclusions



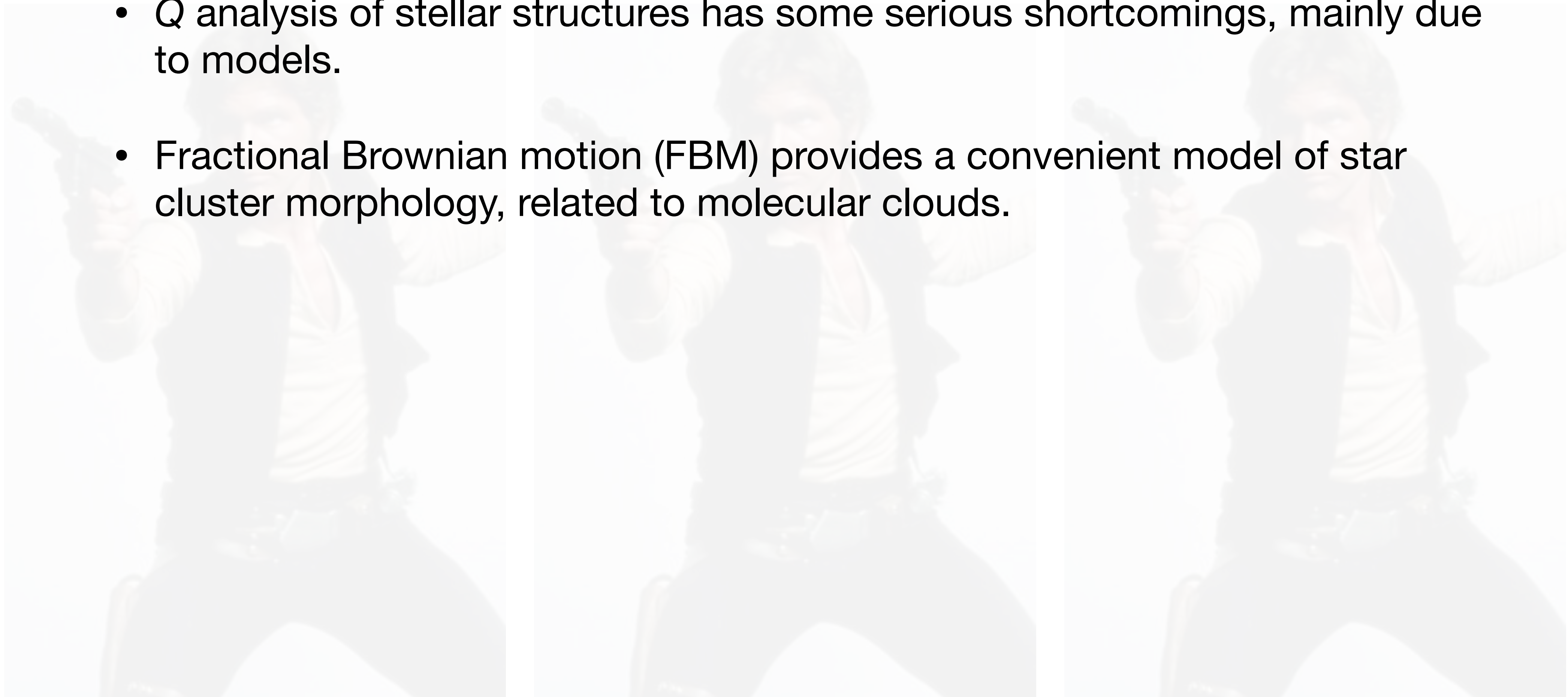
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Article: [Lomax et al., 2018, MNRAS, 480, 371](#)

arXiv: [1804.06844](#)

GitHub: github.com/odlomax/clusterfrac