Towards a Characterization of the Metric Dimension of Barabási-Albert Random Graphs

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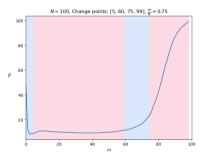
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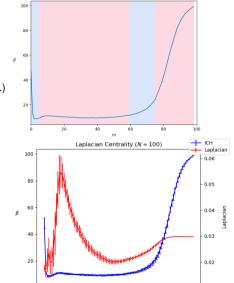
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N = 100, Change points: [5, 60, 75, 99], $\frac{m}{\pi} \approx 0.75$

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- Future work
 - Use more analytical/probabilistic approach
 - Determine upper-bound of metric dimension with high probability
 - Classify regimes of m-values where phase transitions occur

