Focus on: Monte Carlo Tests

Modelling Complex Systems

Q: Is the test statistic on our network, t^* , expected if network is drawn from the null distribution.

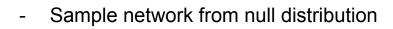
- Only need to be able to sample from the null distribution
- Works a bit like a proof by contradiction start with a hypothesis (that you think is false), and get to a situation which would be unlikely if the hypothesis is true.

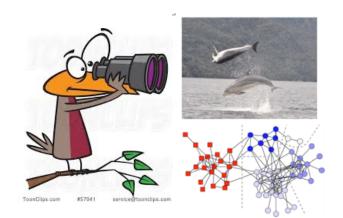
 H_0 - hypothesis : - e.g. that observed network an observation from null distribution

Want to say - observed network not observation from null distribution

- allow some error







- Observed Network

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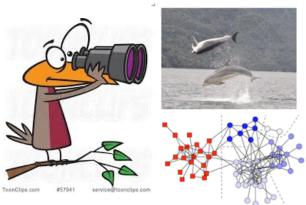


- Sample network from null distribution

- Observed Network
- H_1 reject hypothesis e.g. that observed network not observation from null distribution



- Sample network from null distribution



- Observed Network

If null hypothesis true all orderings of data are equally likely, the probability that the one you observe is among the top m is m/(n + 1)

Also called 'parametric bootstrapping', 'conditional uniform graph test'

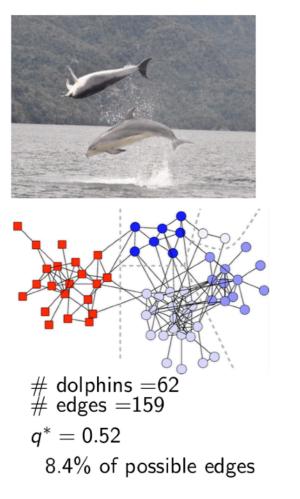
Method for $\alpha = m/(n+1)$

- sample *n* from the null distribution and calculate test statistic $t_1, \ldots t_n$, e.g. sample *n* configuration models and calculate modularity score of each

- order $t^*, t_1, ..., t_n$
- if t* among top m values reject null hypothesis for distribution of network
- rule of thumb, take *m* at least 5.

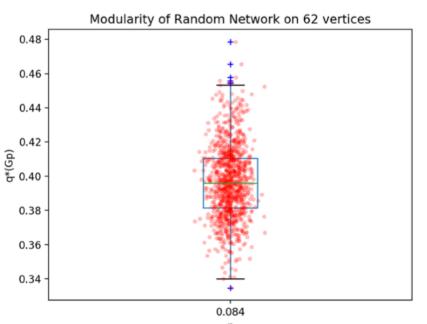
Monte Carlo example

- with Erdos-Renyi random graph as null model.



- Calculate *p* so that null model has same expected degree: $p = #edges / {\frac{#dolphins}{2}}$

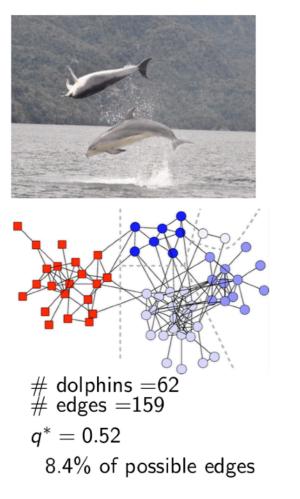
 $q^*(dolphins) > q^*(random network)??$



- The test statistic on our network is $t^* = 0.52$
- Red dots on graph above give test statistic on graphs sampled from the null model. Note t^* is greater than value of test statistic on any generated graph.

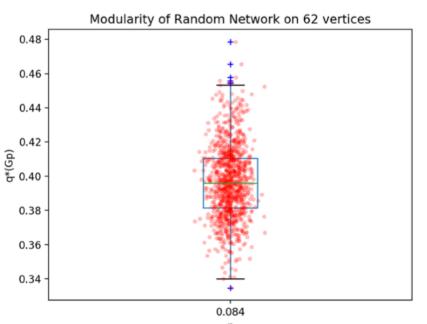
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- e.g. modularity of network compare network to configuration model same degree

- useful when we don't know the distribution
- need to be able to sample from the null distribution
- discrete data has ties (break randomly and method still valid).

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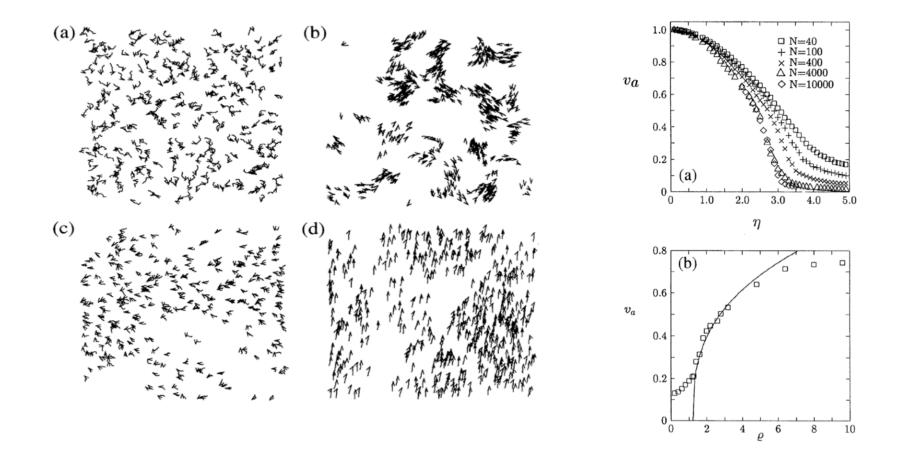
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Watch

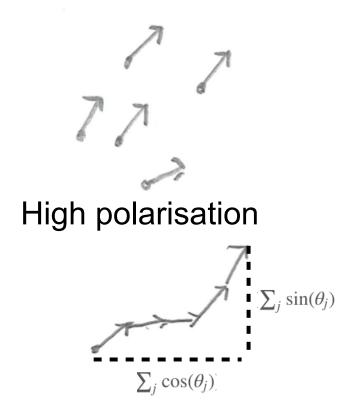
- https://www.youtube.com/watch?v=QT2xj9k00q0
- 1:07-1:11 discusses Monte-Carlo test

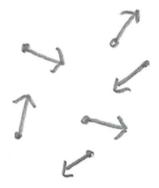
Vicsek Model



Vicsek et al., PRL 75 (1995)

Measure of Alignment: Polarisation





Low Polarisation



Polarisation of: $\theta_1, \theta_2, \dots, \theta_N$

$$= \frac{1}{N} \sqrt{(\sum_j \sin(\theta_j))^2 + (\sum_j \cos(\theta_j))^2}$$