

Is Sharing Caring?

How CRISPR-Cas systems affect rates of Horizontal Gene Transfer

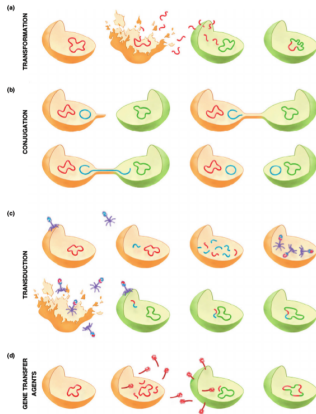
Siddharth Reed

Golding Lab, McMaster University

February 18, 2022

Background

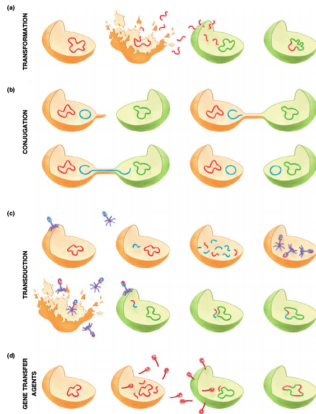
What is Horizontal Gene Transfer?



► Transformation

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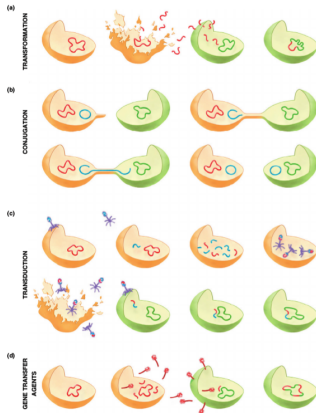
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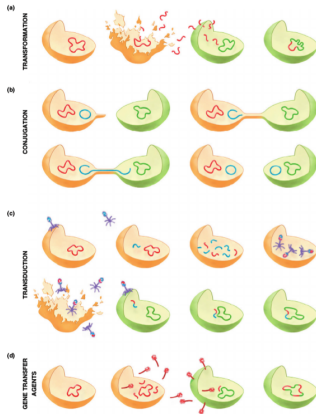


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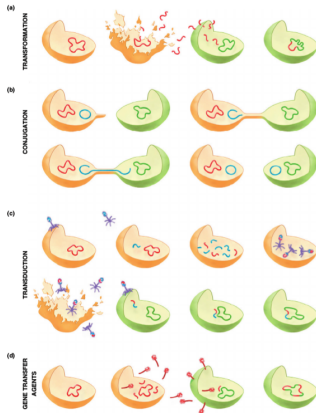


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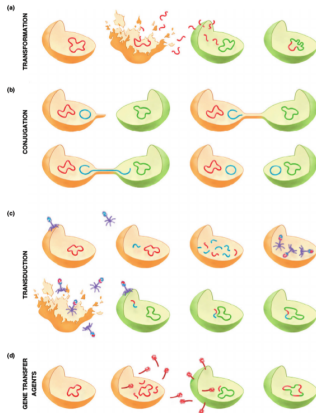


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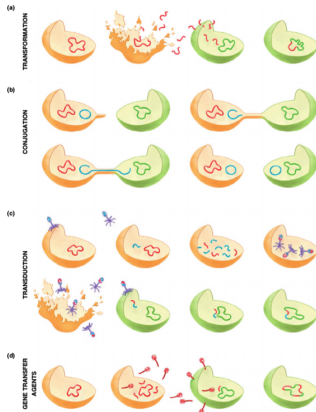


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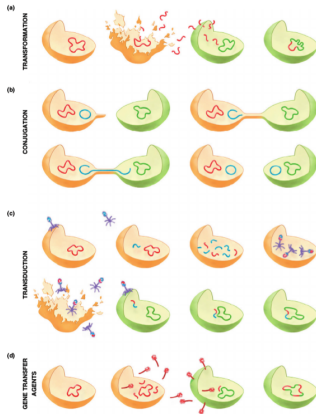


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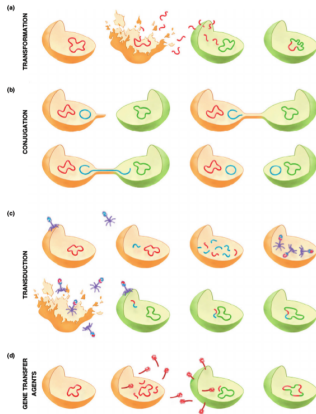


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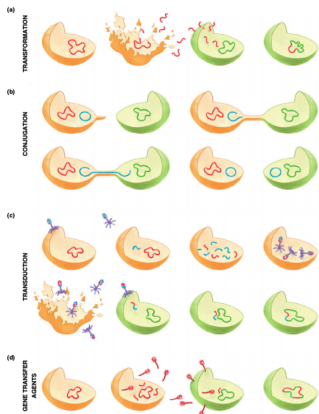


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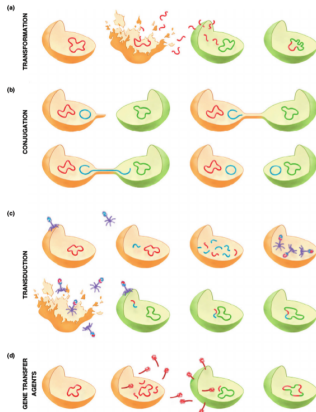


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What is A Network?

- Useful mathematical abstraction of real world system

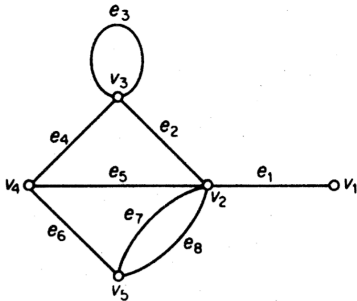


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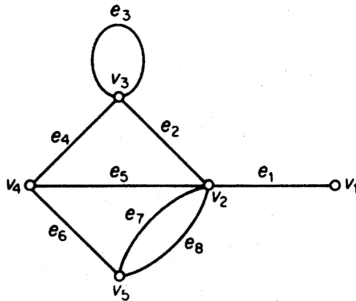


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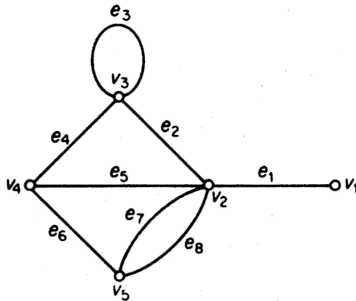


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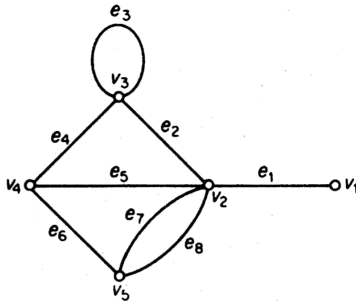


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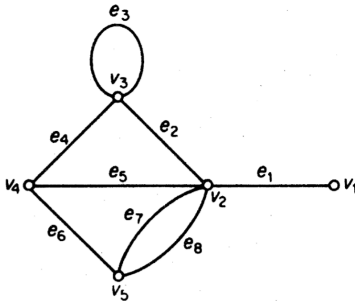


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- ▶ **Nodes are OTUs, edges are inferred HGT rates**

What are CRISPR-Cas systems?

- Adaptive immune system in bacteria

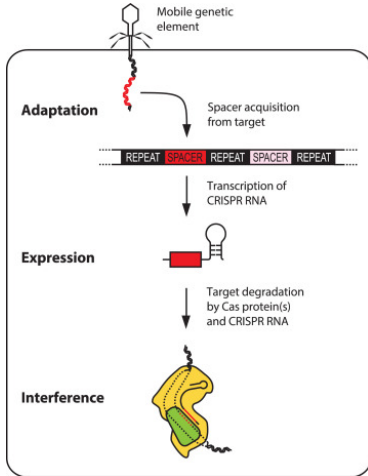
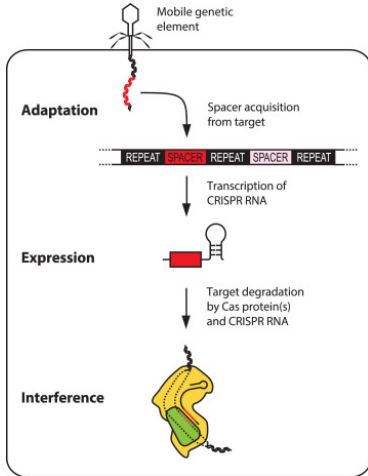


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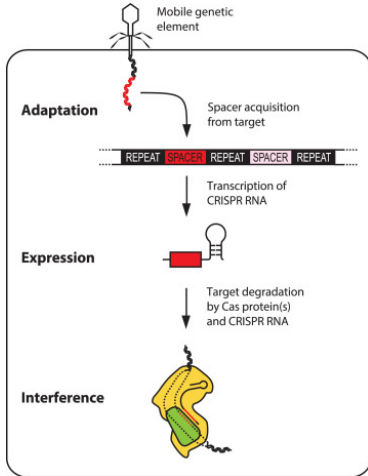
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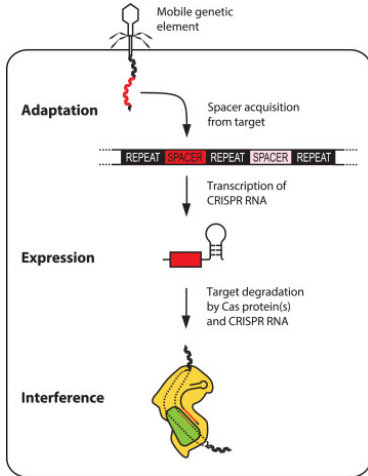
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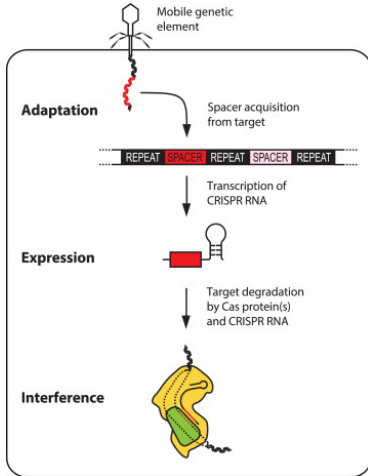
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- ▶ 45% of bacteria have CRISPR loci ($n = 6782$) [4]

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Do CRISPR Systems Affect Horizontal Gene Transfer?

Yes

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 - ◇ Can see inhibitory effects of CRISPR on HGT over short evolutionary time scales
 - ◇ Higher gene indel rates for CRISPR containing OTUs than non-CRISPR containing outgroups

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Fitness Cost Factors

- ▶ Metabolic maintenance [3]

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- ▶ CRISPR can enhance transduction-mediated HGT [9]

My Project

Goals

Within Network Comparisons

For genera with CRISPR containing OTUs, compare the node statistics of CRISPR containing OTUs to non-CRISPR containing OTUs.

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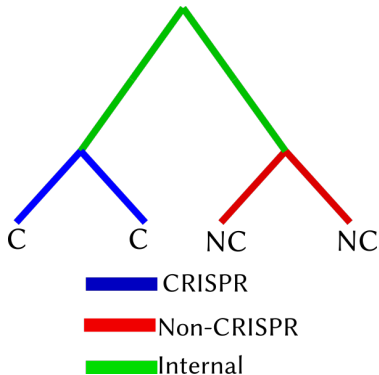
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Gene Indel Rates vs. Network Statistics

Compare gene Indel rates to node/network statistics for CRISPR containing and non-CRISPR containing OTUs

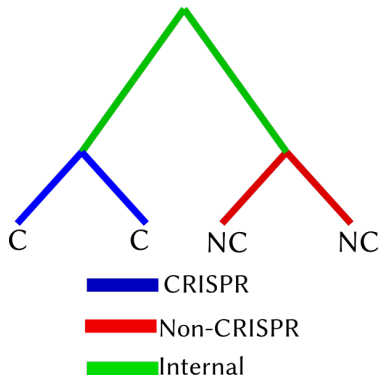
InDel Rate Estimates



Markophylo [11]

- Input: a presence/absence matrix of gene families + species tree

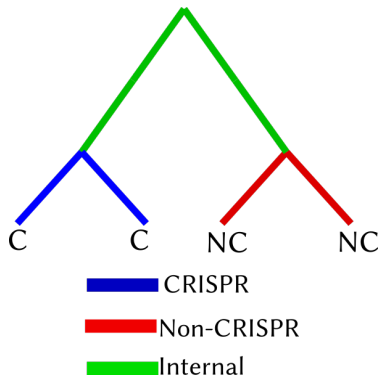
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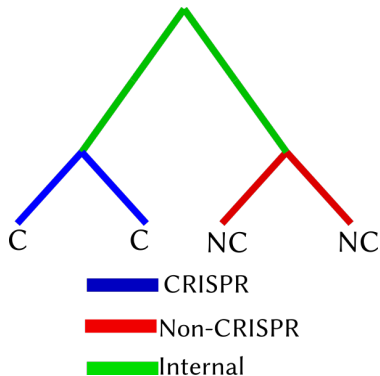
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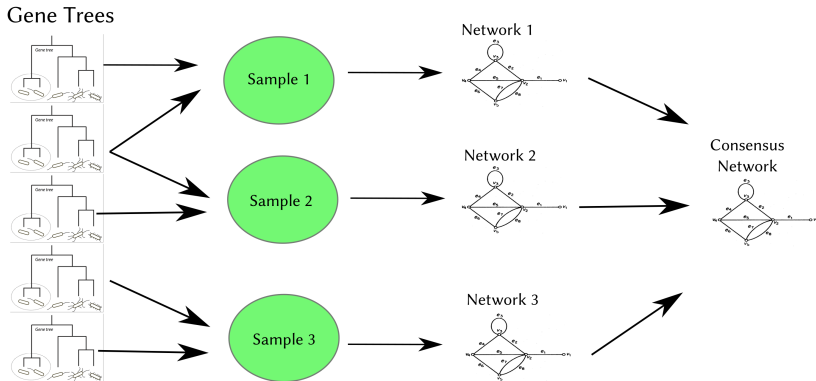
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- ▶ Gene birth-death rates are estimated for each branch partition

Network Sampling



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- ▶ **Network Modularity:** $Q = \frac{1}{2m} \sum_{uv} [W_{uv} - \frac{k_u k_v}{2m}] \delta(u, v)$ where m is the total edge weight, k_u is the degree of u and $\delta(u, v)$ is 1 if u and v both have or do not have CRISPR systems and 0 otherwise. $Q \in [-1, 1]$ [14]

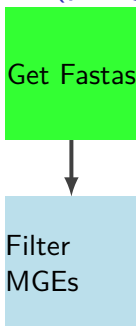
Workflow (per genus)

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Get Fastas

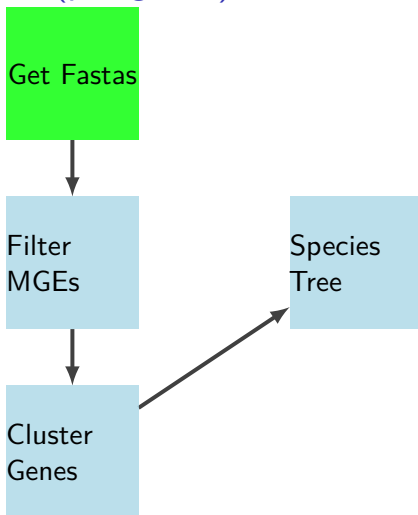
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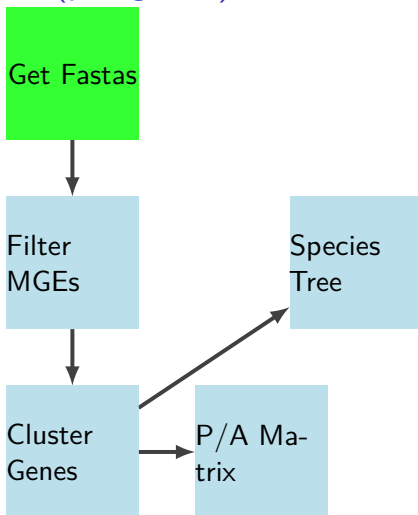
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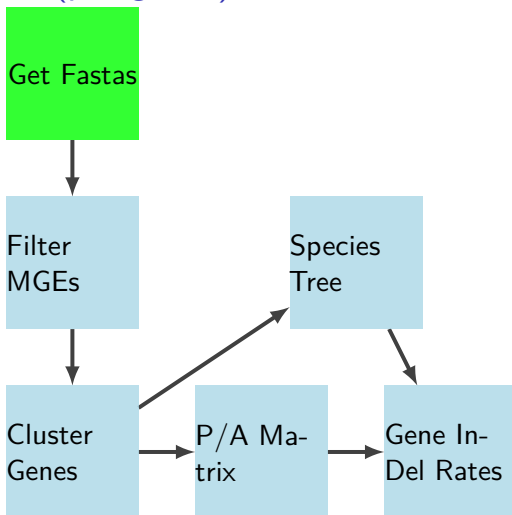
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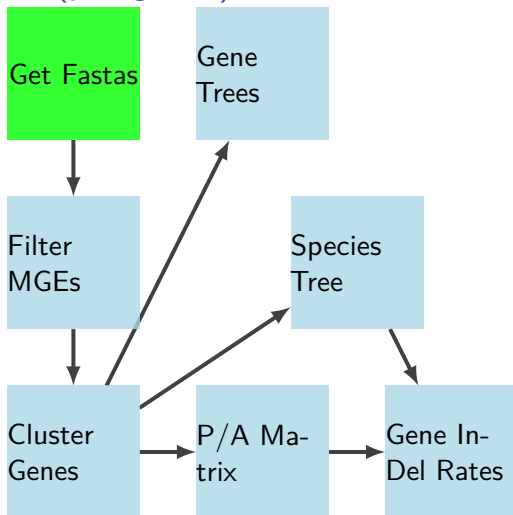
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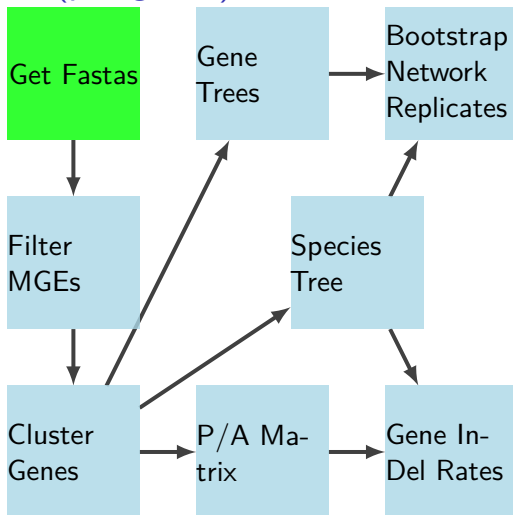
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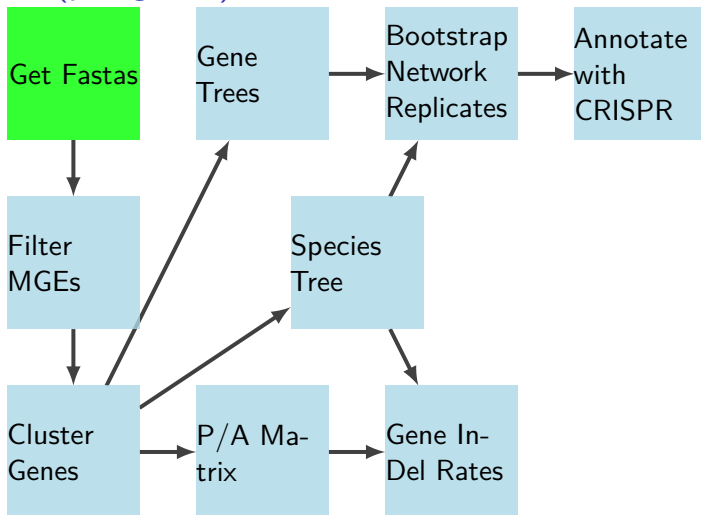
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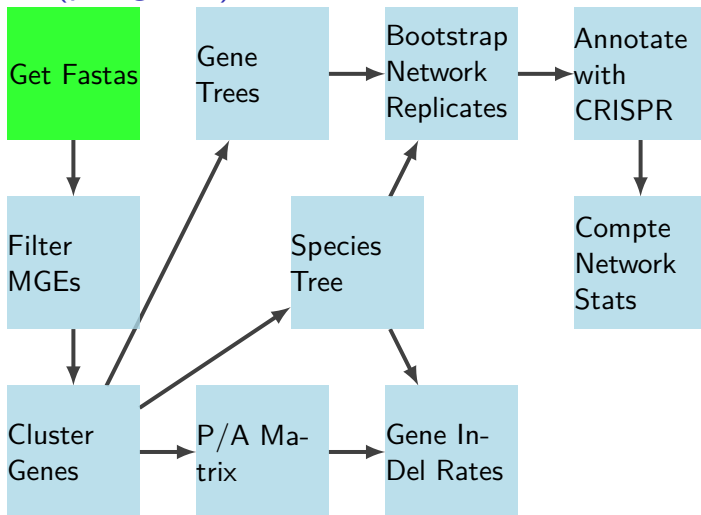
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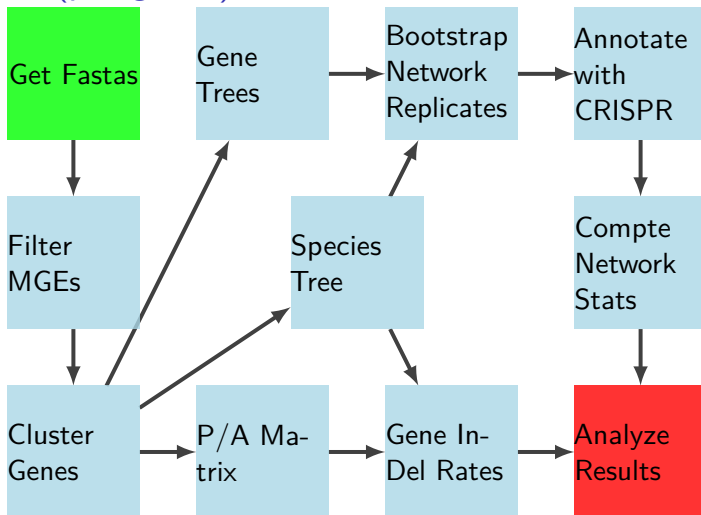
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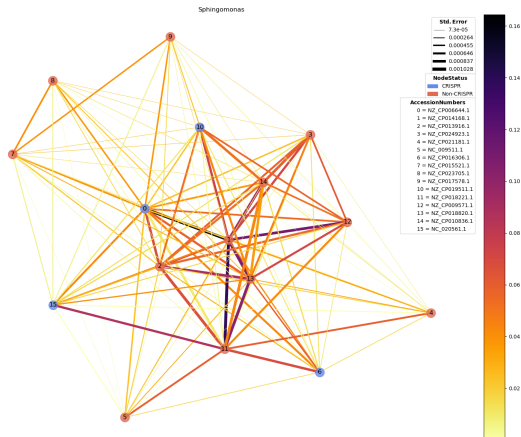


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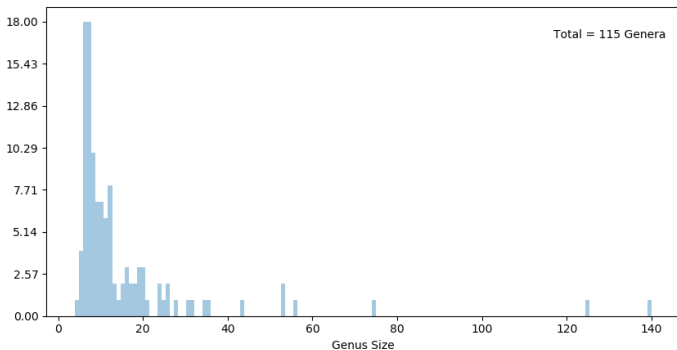


Results

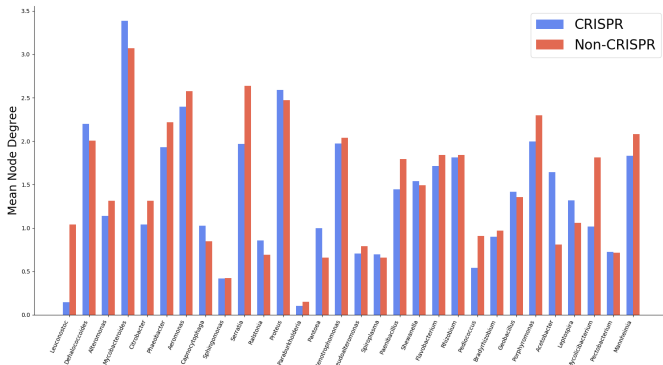
Example “Consensus” Network



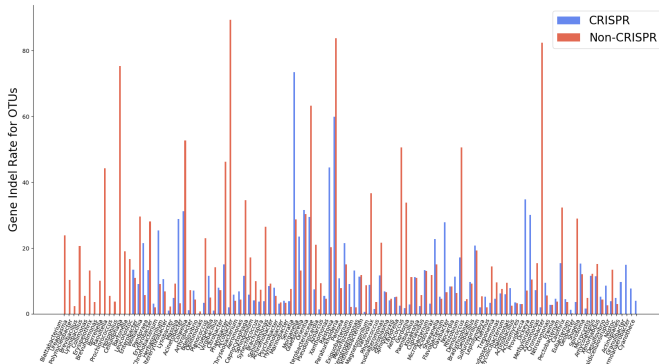
Genus Size Distribution

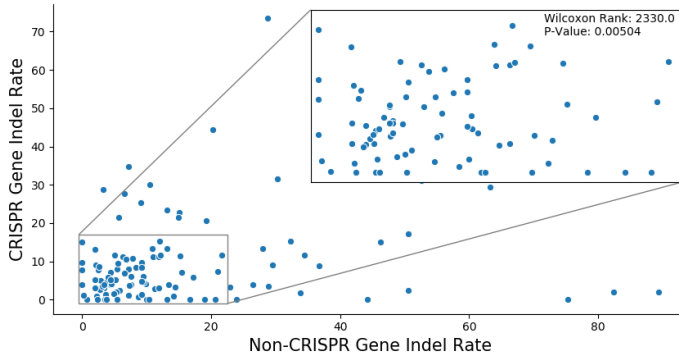


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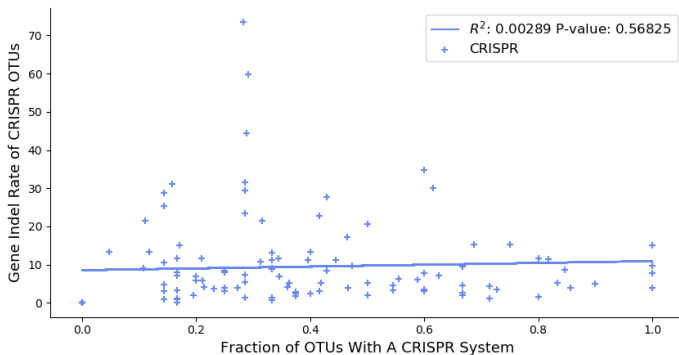


Gene Indel Rates

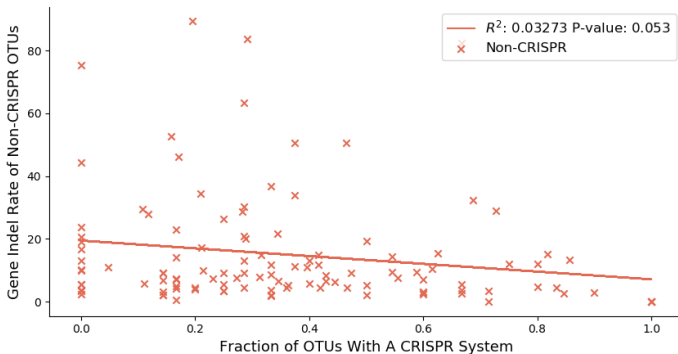




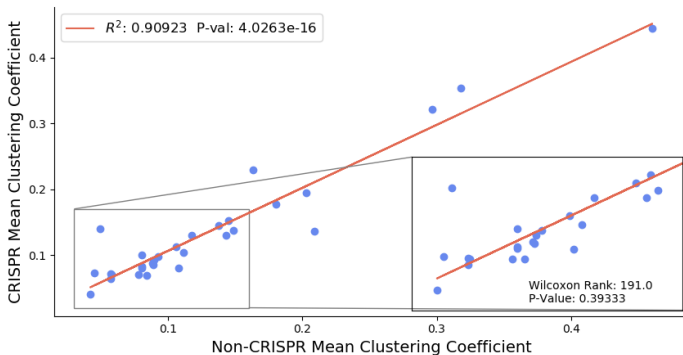
Gene Indel Rate Vs. Fraction of CRISPR OTUs



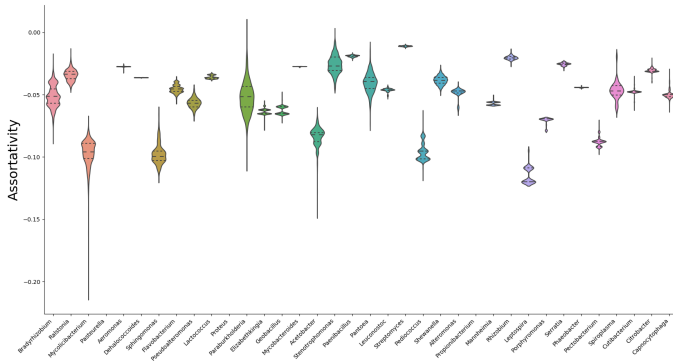
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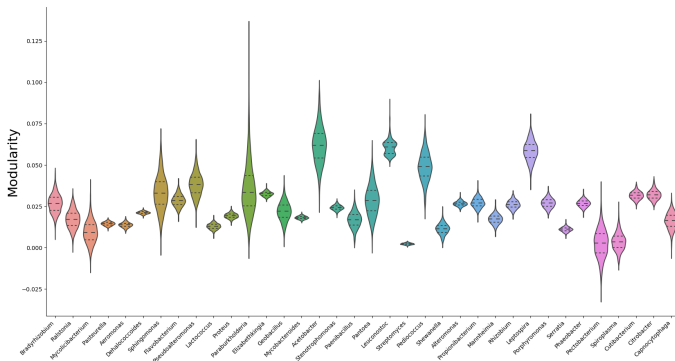
Mean Node Weighted Clustering Coefficient



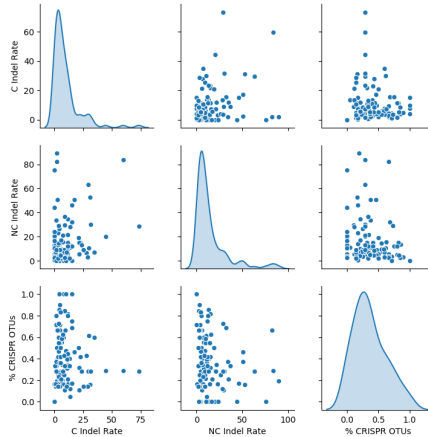
Assortativity Distributions



Modularity Distributions



Indel Rate Pair Plot



Moral of the Study

Findings

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- ▶ CRISPR systems ~ associated with lower HGT rates
 - ◇ Prominent exceptions exist
 - ◇ High mixing between CRISPR and non-CRISPR OTUs
- ▶ Population level effects of CRISPR systems may decrease HGT rates
- ▶ Interplay of CRISPR systems and HGT is complex and warrants further study

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Conclusion

Is Sharing Caring?

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- ▶ Yes, for researchers
- ▶ Jury's still out for bacteria

Thanks

Thank you to

- ▶ Dr. G. Brian Golding
- ▶ Dr. Ben Evans
- ▶ The Golding lab
 - ◇ Caitlin Simopoulos
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 - ◇ Sam Long
 - ◇ George Long
 - ◇ Lucy Zhang
 - ◇ Brianne Laverty
 - ◇ Nicole Zhang
- ▶ Everyone here for listening



Code Availability

All code written for this project is available at https://github.com/DJSiddharthVader/Undergrad_Thesis

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