

Subs Paper Things to Do:

- causes for weird selection and subs results in *Streptomyces*
  - see how often class 4 arises in strep to see what is going on in later portion of the genome (to see if annotation is really a problem)
  - split up the strep data into core and non core and see if results are the same
- ~~make graphs proportional to length of respective cod/non-cod regions~~
- ~~test examples for genes near and far from terminus (robust log reg/results)~~
- ~~linear regression on 10kb regions for weighted and non-weighted substitutions~~
- ~~average number of substitutions in 20kb regions near and far from the origin~~
- ~~figure out why the data is weird for number of cod/non-cod sites~~
- why are the lin reg of  $dN$ ,  $dS$  and  $\omega$  NS but the subs graphs are...explain!
- grey out outliers in subs graphs?
- mol clock for my analysis?
- GC content? COG? where do these fit?

Gene Expression Paper Things to Do:

- ~~linear regression on 10kb regions~~
- ~~put new 10kb lin reg and # of genes over 10kb lin reg into paper~~
- ~~write about  $\uparrow$  in methods and discussion~~
- ~~put expression lin reg and # coding sites log reg into supplement~~
- ~~write about  $\uparrow$  in paper and how results are the same~~
- ~~update supplementary figures/file~~
- ~~correlation of gene expression across strains~~
  - ~~make graphs pretty and more informative with label names~~
  - ~~add them to supplement with a mini write up of what we did and why~~
  - ~~mention this in the actual paper~~
- if necessary add a phylogenetic component to the analysis
- potentially remove genes that have been recently translocated from the analysis
- model gene exp + position + number of genes

- split up the strep data into core and non core and see if results are the same
- what is going on with *Streptomyces* number of genes changing drastically from core to non-core
- codon bias?
- what is going on with really high gene expression bars
- edit paper
- submit paper

#### Inversions and Gene Expression Letter Things to Do:

- ~~check if opposite strand in progressiveMauve means an inversions (check visual matches the xmfa)~~
- ~~check if PARSNP and progressiveMauve both identify the same inversions (check xmfa file)~~
- create latex template for paper
- ~~put notes from papers into doc~~
- ~~use large PARSNP alignment to identify inversions~~
- confirm inversions with dot plot
- make dot plot of just gene presence and absence matrix (instead of each site) to see if this will go better
- look up inversions and small RNA's paper Marie was talking about at Committee meeting
- write outline for letter
- write Abstract
- write intro
- write methods
- compile tables (supplementary)
- write results
- write discussion
- write conclusion
- do same ancestral/phylogenetic analysis that I did in the subs paper

#### General Things to Do:

- summarize references 40 and 56 from Committee meeting report (Brian was asking)
- read and make notes on papers I found for dissertation intro

## Last Week

- ✓compaired trimal and Gblocks
- ✓look into other alignment assessment programs
- ✓how do algorithms for Gblocks and trimal work?
- ✓min length for aligned segments
- ✓get pipeline for implementing Gblocks (additional steps)
- ✓write code to parse the Gblocks alignment and print out only the “good” sections of the aln
- ✓more *Streptomyces* genomes?
- ✓other linear species?

**Alignment Assessment** It looks like Gblocks and trimal do very similar things. Trimal will give a score for each site in the alignment and can also calculate this score over a sliding window. However, it seems like it will be hard to implement into my pipeline because the output file is not in an easily parsible format. I decided to go with Gblocks because it will be easy to implement.

I looked into other alignment assesement programs and like I mentioned to you, the only ones that use a tree, re-align the sequences themselves. The idea is that you need an alignment to make a tree so the ones that assess the alignment but do not re-align are trying to be implemented even before you build a tree. We decided that not re-aligning with another program is best.

As I mentioned, trimal will calculate scores for each site or over a specified sliding window. It can calculated different scores, someusing the proportion of sequences with say a gap or the same base, and it also uses an identity matrix to calculate distances and then weights the score based on those. Gblocks is completely proportional. So x number of sequences need to have the same base for it to be considered “conserved”. It also has cutoffs for the “highly conserved” category.

As discussed, the short segments of the alignments are often inaccurate and not comparing homologous regions. So we decided to make each segment a minimum of 100bp long. This means that my code will go through the mafft alignment first, making sure that segments are comparing codon 1 with codon 1...etc. And then those sections will be passed through Gblocks to ensure that they are min 100bp and well aligned.

**Implementation of Gblocks** I tested out how to implement Gblocks into my current pipeline and worked out the kinks. It is really quick (seconds to assess all blocks) and can be implemented for all bacteria in both coding and non-coding. I also wrote a short code to parse the Gblocks output and print out the “good” segments of the alignment (python).

***Streptomyces* Genomes** As I mentioned to you this week, I was looking into other *Streptomyces* genomes that we could potentially use and it looks like there are 2 options for the analysis:

1. only use 3 taxa, 2 strains of *S. lividans* and 1 of *S. coelicolor*, these have **6 blocks total**
2. only use 4 strains of *S. venezuelae*, these have **2 blocks total**

So we need to decide if we want more rearrangements, or more taxa. I still need to think about this and decide.

**Other Linear Genomes** I looked into other bacteria with at least one linear chromosome and it looks like *A. tumefaciens* has a circular and a linear chrom, and *Borrelia burgdorferi* has one linear chromosome. *A. tumefaciens* has 15 complete ref genomes, and *Borrelia burgdorferi* has 10 complete ref genomes. I aligned these last week with progressiveMauve in multiple groupings to see if there are any that are similar enough but not too similar. I will report back on the results when they are done.

## This Week

I would like to switch and work on the gene expression paper again.

1. phylogenetic analysis with gene expression in *E. coli*?
2. remove genes that have been recently translocated from analysis?
3. model gene expression + position + number of genes
4. split up *Streptomyces* data into core and non-core and see if the results are the same (do same for number of genes)

## Next Week

Back to the substitutions project:

1. pick which group of *Streptomyces* genomes to use
2. what to do about length cut off for the non-coding regions? am I putting this whole section in the supplement?
3. assess the *Borrelia burgdorferi* mauve plots for similar seqs
4. align the *A. tumefaciens* taxa into more specific phylo groups with mauve
5. assess the agro mauve plots

Bacteria and Replicon	Near Origin			Near Terminus		
	$dN$	$dS$	$\omega$	$dN$	$dS$	$\omega$
<i>E. coli</i> Chromosome	NS	NS	NS	NS	NS	NS
<i>B. subtilis</i> Chromosome	NS	NS	NS	NS	NS	NS
<i>Streptomyces</i> Chromosome	—	—	—	—	—	—
<i>S. meliloti</i> Chromosome	$3.77 \times 10^{-8**}$	$3.54 \times 10^{-7**}$	$1.23 \times 10^{-6**}$	NS	NS	NS
<i>S. meliloti</i> pSymA	NS	NS	$3.42 \times 10^{-5*}$	NS	NS	NS
<i>S. meliloti</i> pSymB	NS	NS	NS	$-3.24 \times 10^{-7**}$	$8.33 \times 10^{-6***}$	NS

Table 1: Linear regression for  $dN$ ,  $dS$ , and  $\omega$  calculated for each bacterial replicon for the 20 genes closest and 20 genes farthest from the origin of replication. All results are marked with significance codes as followed:  $p: < 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .

Bacteria and Replicon	Protein Coding Sequences	Non-Protein Coding Sequences
<i>E. coli</i> Chromosome	$-4.308 \times 10^{-8***}$	NS
<i>B. subtilis</i> Chromosome	$-4.971 \times 10^{-8***}$	$-1.055 \times 10^{-7***}$
<i>Streptomyces</i> Chromosome		
<i>S. meliloti</i> Chromosome	$-1.903 \times 10^{-7***}$	$-2.900 \times 10^{-7***}$
<i>S. meliloti</i> pSymA	$-6.642 \times 10^{-7***}$	$-1.263 \times 10^{-6***}$
<i>S. meliloti</i> pSymB	$1.769 \times 10^{-7***}$	$4.771 \times 10^{-7***}$

Table 2: Logistic regression analysis of the number of substitutions along all positions of the genome of the respective bacteria replicons. These genomic positions were split up into the coding and non-coding regions of the genome. Grey coloured boxes indicate a negative logistic regression coefficient estimate. All results are statistically significant. Logistic regression was calculated after the origin of replication was moved to the beginning of the genome and all subsequent positions were scaled around the origin accounting for bidirectionality of replication. All results are marked with significance codes as followed:  $< 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .

Bacteria and Replicon	Protein Coding				Non-Protein Coding			
	Correlation Coefficient 20kb Near		Number of Substitutions per 20kb Near		Correlation Coefficient 20kb Near		Number of Substitutions per 20kb Near	
	Origin	Terminus	Origin	Terminus	Origin	Terminus	Origin	Terminus
<i>E. coli</i> Chromosome	$-2.889 \times 10^{-5*}$	NS	$2.87 \times 10^{-2}$	$4.24 \times 10^{-2}$	$-4.316 \times 10^{-5***}$	$-8.209 \times 10^{-5*}$	$1.095 \times 10^{-2}$	$4.45 \times 10^{-3}$
<i>B. subtilis</i> Chromosome	NS	$1.863 \times 10^{-5*}$	$4.8 \times 10^{-3}$	$3.06 \times 10^{-2}$	$1.017 \times 10^{-4*}$	$5.823 \times 10^{-5***}$	$8 \times 10^{-4}$	$6.75 \times 10^{-3}$
<i>Streptomyces</i> Chromosome								
<i>S. meliloti</i> Chromosome	NS	NS	$4.05 \times 10^{-3}$	$2 \times 10^{-4}$	NS	NS	$9 \times 10^{-4}$	$1.5 \times 10^{-4}$
<i>S. meliloti</i> pSymA	NS	NS	$6.15 \times 10^{-3}$	$1.9 \times 10^{-3}$	$1.403 \times 10^{-4***}$	$-2.220 \times 10^{-4**}$	$2.8 \times 10^{-3}$	$5.5 \times 10^{-4}$
<i>S. meliloti</i> pSymB	$-1.553 \times 10^{-5*}$	$-4.908 \times 10^{-5***}$	$3.23 \times 10^{-2}$	$2.36 \times 10^{-2}$	NS	$-4.557 \times 10^{-5**}$	$5.1 \times 10^{-3}$	$5.4 \times 10^{-3}$

Table 3: Logistic regression on 20kb closest and farthest from the origin of replication after accounting for bidirectional replication and outliers. All results are marked with significance codes as followed:  $< 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .

Bacteria and Replicon	Protein Coding		Non-Protein Coding	
	Weighted	Non-Weighted	Weighted	Non-Weighted
<i>E. coli</i> Chromosome	$-4.87 \times 10^{-10**}$	$-1.839 \times 10^{-4***}$	NS	$-2.244 \times 10^{-5***}$
<i>B. subtilis</i> Chromosome	NS	$-2.031 \times 10^{-4**}$	NS	$-2.885 \times 10^{-5**}$
<i>Streptomyces</i> Chromosome				
<i>S. meliloti</i> Chromosome	$-1.341 \times 10^{-10**}$	$-1.461 \times 10^{-5**}$	$-3.490 \times 10^{-10*}$	NS
<i>S. meliloti</i> pSymA	NS	NS	$-1.144 \times 10^{-8**}$	$-6.74 \times 10^{-5**}$
<i>S. meliloti</i> pSymB	NS	NS	NS	NS

Table 4: Linear regression on 10kb sections of the genome with increasing distance from the origin of replication after accounting for bidirectional replication. Weighted columns have the total number of substitutions in each 10kb section of the genome divided by the total number of protein coding and non-protein coding sites in the genome. Non-weighted columns are performing a linear regression on the total number of substitutions in each 10kb section of the genome. All results are marked with significance codes as followed:  $< 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .

Bacteria and Replicon	Gene Expression 10kb
<i>E. coli</i> Chromosome	$-2.742 \times 10^{-5**}$
<i>B. subtilis</i> Chromosome	$-2.198 \times 10^{-5*}$
<i>Streptomyces</i> Chromosome	$-5.230 \times 10^{-7***}$
<i>S. meliloti</i> Chromosome	NS
<i>S. meliloti</i> pSymA	NS
<i>S. meliloti</i> pSymB	NS

Table 5: Linear regression analysis of the median counts per million expression data for 10kb segments of the genome of the respective bacteria replicons. Linear regression was calculated after the origin of replication was moved to the beginning of the genome and all subsequent positions were scaled around the origin accounting for bidirectionality of replication. All results are marked with significance codes as followed:  $< 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .

Bacteria and Replicon	Coefficient Estimate	Standard Error	P-value
<i>E. coli</i> Chromosome	$-6.03 \times 10^{-5}$	$1.28 \times 10^{-5}$	$2.8 \times 10^{-6}$
<i>B. subtilis</i> Chromosome	$-9.7 \times 10^{-5}$	$2.0 \times 10^{-5}$	$1.2 \times 10^{-6}$
<i>Streptomyces</i> Chromosome	$-1.17 \times 10^{-6}$	$1.04 \times 10^{-7}$	$< 2 \times 10^{-16}$
<i>S. meliloti</i> Chromosome	$3.97 \times 10^{-5}$	$4.25 \times 10^{-5}$	NS ( $3.5 \times 10^{-1}$ )
<i>S. meliloti</i> pSymA	$1.39 \times 10^{-3}$	$2.53 \times 10^{-4}$	$4.9 \times 10^{-8}$
<i>S. meliloti</i> pSymB	$1.46 \times 10^{-4}$	$2.03 \times 10^{-4}$	NS ( $5.34.7 \times 10^{-1}$ )

Table 6: Linear regression analysis of the median counts per million expression data along the genome of the respective bacteria replicons. Linear regression was calculated after the origin of replication was moved to the beginning of the genome and all subsequent positions were scaled around the origin accounting for bidirectionality of replication.

Bacteria and Replicon	Coefficient Estimate
<i>E. coli</i> Chromosome	NS
<i>B. subtilis</i> Chromosome	$-2.682 \times 10^{-6}***$
<i>Streptomyces</i> Chromosome	$-2.360 \times 10^{-6}***$
<i>S. meliloti</i> Chromosome	$-2.074 \times 10^{-6}***$
<i>S. meliloti</i> pSymA	NS
<i>S. meliloti</i> pSymB	$-4.19 \times 10^{-6}*$

Table 7: Linear regression analysis of the total number of protein coding genes per 10kb along the genome of the respective bacteria replicons. Linear regression was calculated after the origin of replication was moved to the beginning of the genome and all subsequent positions were scaled around the origin accounting for bidirectionality of replication. All results are marked with significance codes as followed:  $< 0.001 = '***'$ ,  $0.001 < 0.01 = '**'$ ,  $0.01 < 0.05 = '*'$ ,  $> 0.05 = 'NS'$ .



Bacteria and Replicon	$dN$	$dS$	$\omega$
<i>E. coli</i> Chromosome	NS	NS	NS
<i>B. subtilis</i> Chromosome	NS	NS	$-9.08 \times 10^{-6*}$
<i>Streptomyces</i> Chromosome	NS	NS	NS
<i>S. meliloti</i> Chromosome	NS	NS	NS
<i>S. meliloti</i> pSymA	NS	NS	NS
<i>S. meliloti</i> pSymB	NS	NS	$1.163 \times 10^{-5*}$

Table 8: Linear regression for  $dN$ ,  $dS$ , and  $\omega$  calculated for each bacterial replicon on a per genome basis. All results are marked with significance codes as followed:  $p < 0.001 = '***'$ ,  $0.001 < p < 0.01 = '**'$ ,  $0.01 < p < 0.05 = '*'$ ,  $p > 0.05 = 'NS'$ .

Bacteria and Replicon	Average Expression Value (CPM)
<i>E. coli</i> Chromosome	160.500
<i>B. subtilis</i> Chromosome	176.400
<i>Streptomyces</i> Chromosome	6.084
<i>S. meliloti</i> Chromosome	271.400
<i>S. meliloti</i> pSymA	690.100
<i>S. meliloti</i> pSymB	595.700

Table 9: Arithmetic gene expression calculated across all genes in each replicon. Expression values are represented in Counts Per Million.

Bacteria and Replicon	Gene Average			Genome Average		
	dS	dN	$\omega$	dS	dN	$\omega$
<i>E. coli</i> Chromosome	1.0468	0.1330	1.3183	0.6491	0.0364	0.2432
<i>B. subtilis</i> Chromosome	4.652	0.2333	2.4200	1.0879	0.0703	0.3852
<i>Streptomyces</i> Chromosome	13.4950	2.0973	21.0423	5.1256	0.8911	8.9146
<i>S. meliloti</i> Chromosome	0.0184	0.0012	0.1069	0.0187	0.0013	0.0962
<i>S. meliloti</i> pSymA	1.0602	0.7451	5.1290	0.4100	0.0863	0.8311
<i>S. meliloti</i> pSymB	3.2602	0.0256	0.3878	0.1436	0.0100	0.1943

Table 10: Weighted averages calculated for each bacterial replicon on a per genome basis using the gene length as the weight. Arithmetic mean calculated for the per gene averages for each bacterial replicon.

Bacteria Strain/Species	GEO Accession Number	Date Accessed
<i>E. coli</i> K12 MG1655	GSE60522	December 20, 2017
<i>E. coli</i> K12 MG1655	GSE73673	December 19, 2017
<i>E. coli</i> K12 MG1655	GSE85914	December 19, 2017
<i>E. coli</i> K12 MG1655	GSE40313	November 21, 2018
<i>E. coli</i> K12 MG1655	GSE114917	November 22, 2018
<i>E. coli</i> K12 MG1655	GSE54199	November 26, 2018
<i>E. coli</i> K12 DH10B	GSE98890	December 19, 2017
<i>E. coli</i> BW25113	GSE73673	December 19, 2017
<i>E. coli</i> BW25113	GSE85914	December 19, 2017
<i>E. coli</i> O157:H7	GSE46120	August 28, 2018
<i>E. coli</i> ATCC 25922	GSE94978	November 23, 2018
<i>B. subtilis</i> 168	GSE104816	December 14, 2017
<i>B. subtilis</i> 168	GSE67058	December 16, 2017
<i>B. subtilis</i> 168	GSE93894	December 15, 2017
<i>B. subtilis</i> 168	GSE80786	November 16, 2018
<i>S. coelicolor</i> A3	GSE57268	March 16, 2018
<i>S. natalensis</i> HW-2	GSE112559	November 15, 2018
<i>S. meliloti</i> 1021 Chromosome	GSE69880	December 12, 2017
<i>S. meliloti</i> 2011 pSymA	NC_020527 (Dr. Finan)	April 4, 2018
<i>S. meliloti</i> 1021 pSymA	GSE69880	November 15, 18
<i>S. meliloti</i> 2011 pSymB	NC_020560 (Dr. Finan)	April 4, 2018
<i>S. meliloti</i> 1021 pSymB	GSE69880	November 15, 18

Table 11: Summary of strains and species found for each gene expression analysis. Gene Expression Omnibus accession numbers and date accessed are provided.