

## K-mer and contaminant analysis



- What is a k-mer?
  - A k-mer is a sequence of nucleotides of length k.
  - Examples of a 6-mer
    - ACGTCT
    - TGACTA
    - GATCCC
- A read of length L has  $L-k+1$  k-mers.

1 read:

100 bp



Kmers:

$k=21\text{bp}$

$$N = (L - k + 1)$$

$$(100\text{bp} - 21\text{ bp} + 1)$$

80

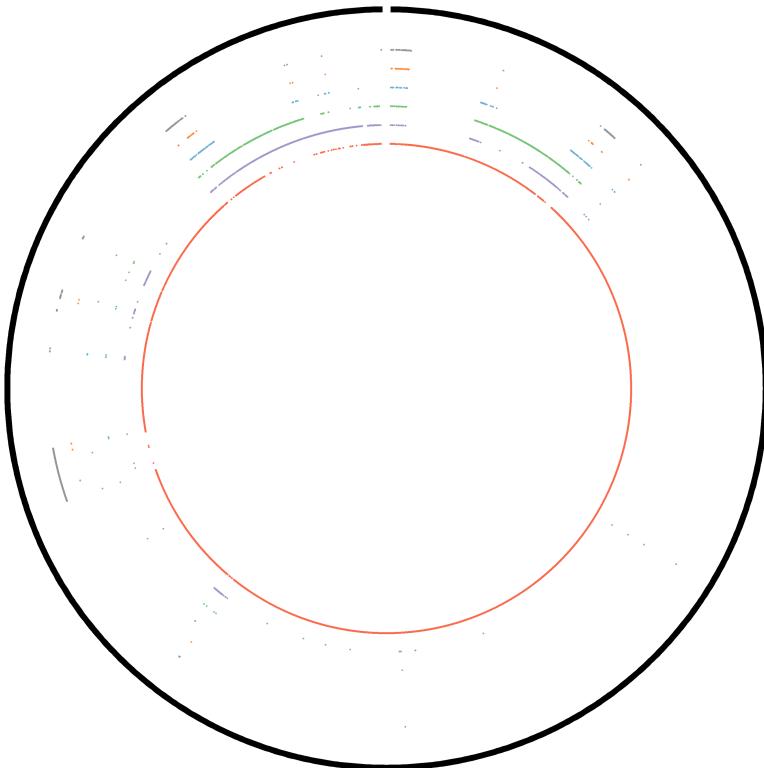


.....



# K-mer Analysis

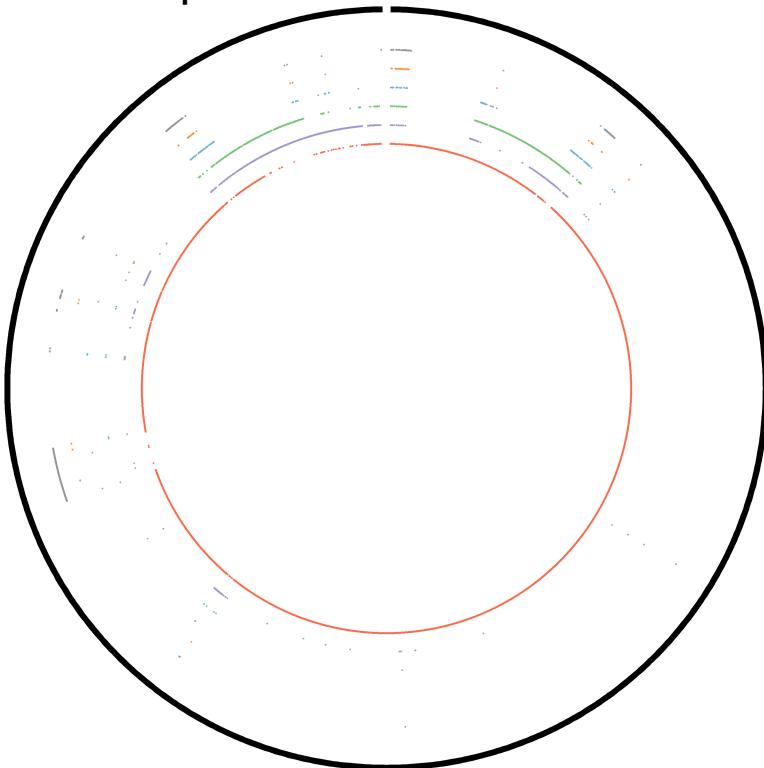
How many k-mers are distinct in your genome?



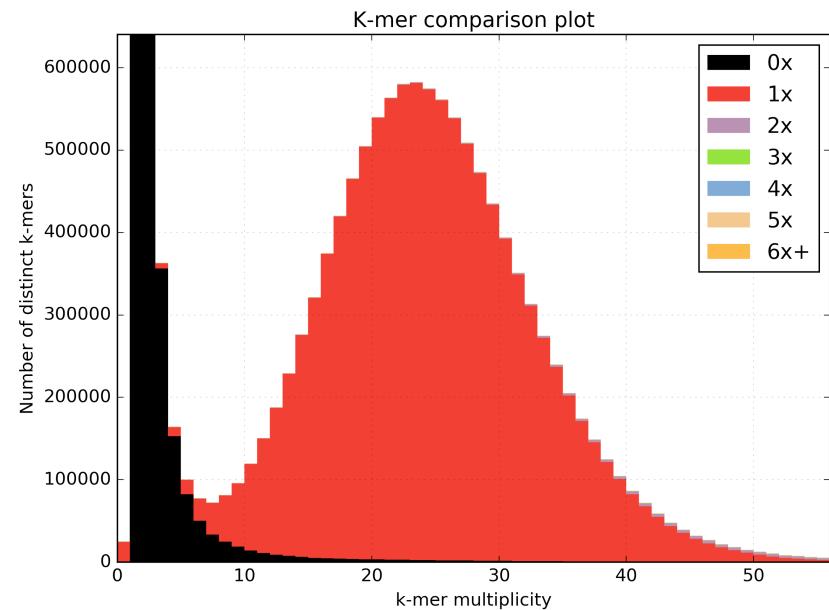
- Chr 1 of yeast strain s288c
  - Haploid
  - Linear, but plotted as circle
  - Black: position along chromosome
  - K-mer frequency in genome
    - K=27
    - Red: 1
    - Purple: 2
    - Green: 3
    - Blue: 4
    - Orange: 5
    - Grey: 6+
  - Majority of k-mers are distinct.

# K-mer Analysis

Frequency in the genome  
Haploid



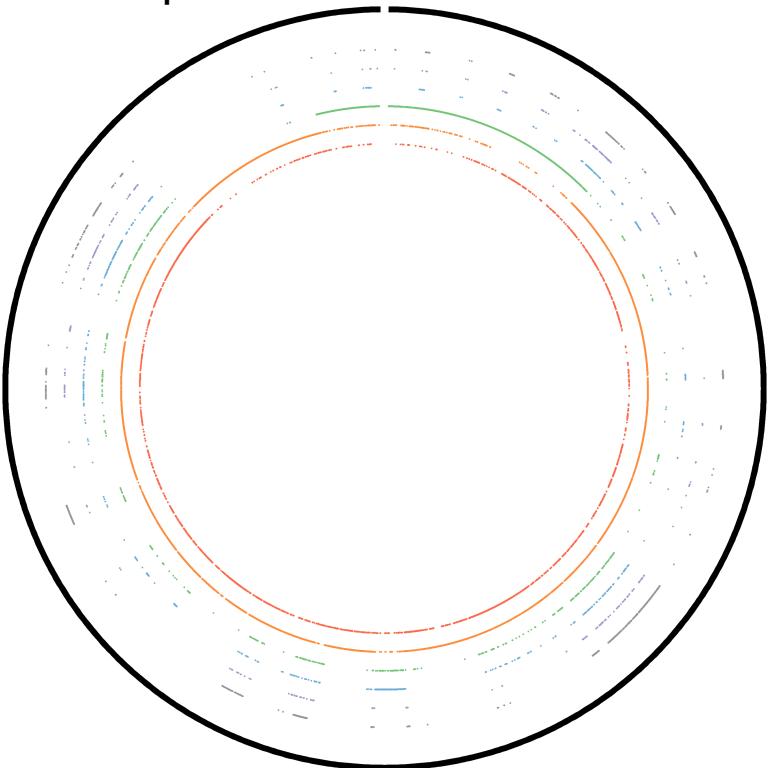
Frequency of those k-mers in  
the sequence data set



x-axis: Frequency of the  
distinct k-mer in the data  
y-axis: Count of distinct  
k-mers with frequency x

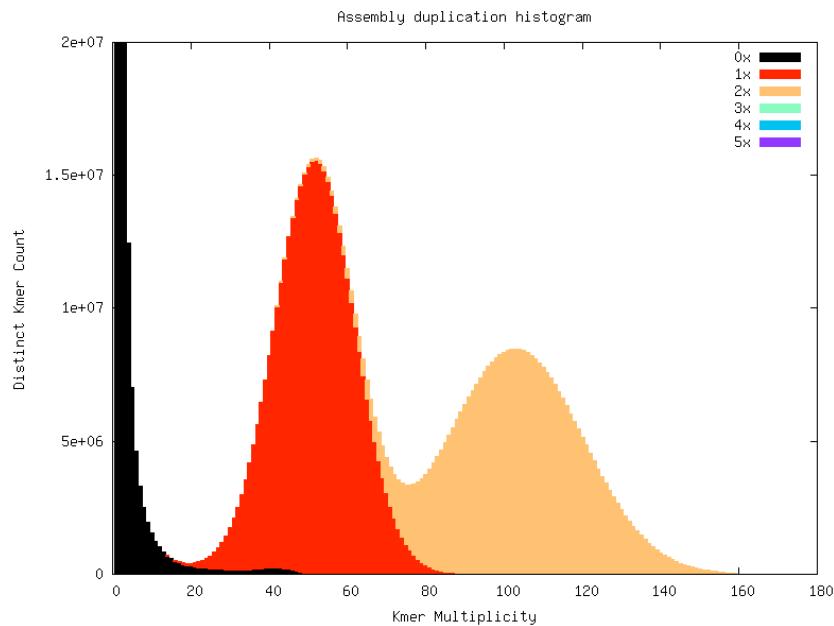
# K-mer Analysis

Frequency in the genome  
Diploid



Contig from A.thaliana phased genome

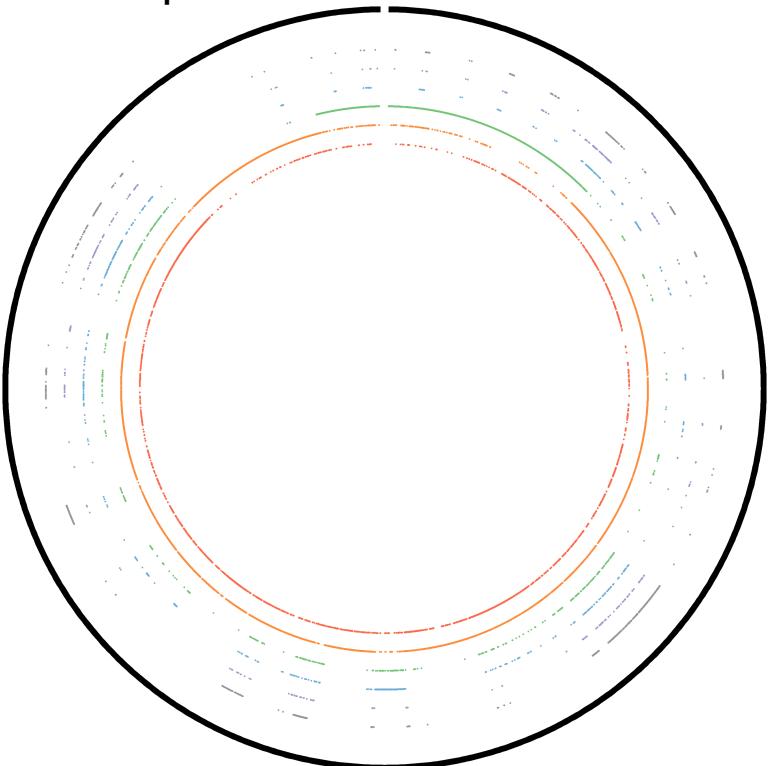
Frequency of those k-mers in  
the sequence data set



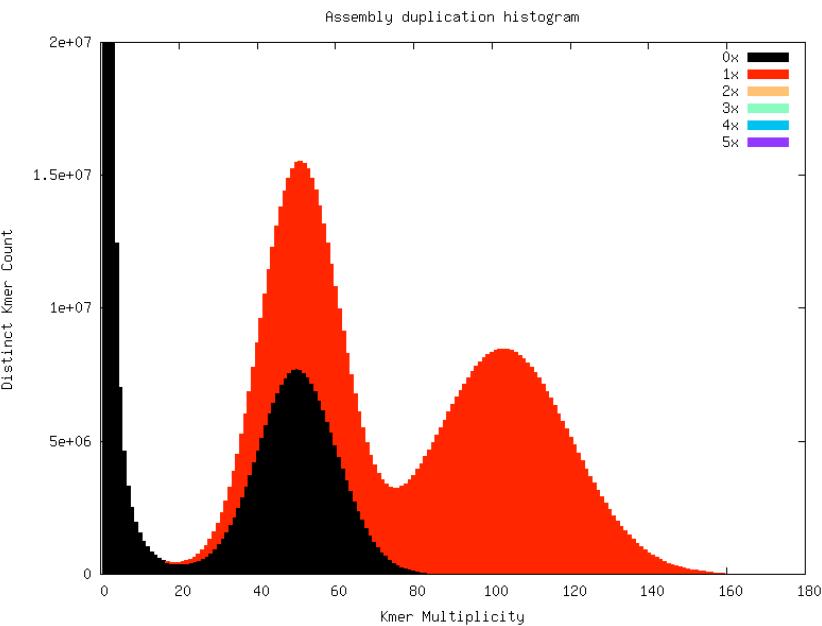
x-axis: Frequency of the  
distinct k-mer in the data  
y-axis: Count of distinct  
k-mers with frequency x

# K-mer Analysis

Frequency in the genome  
Diploid



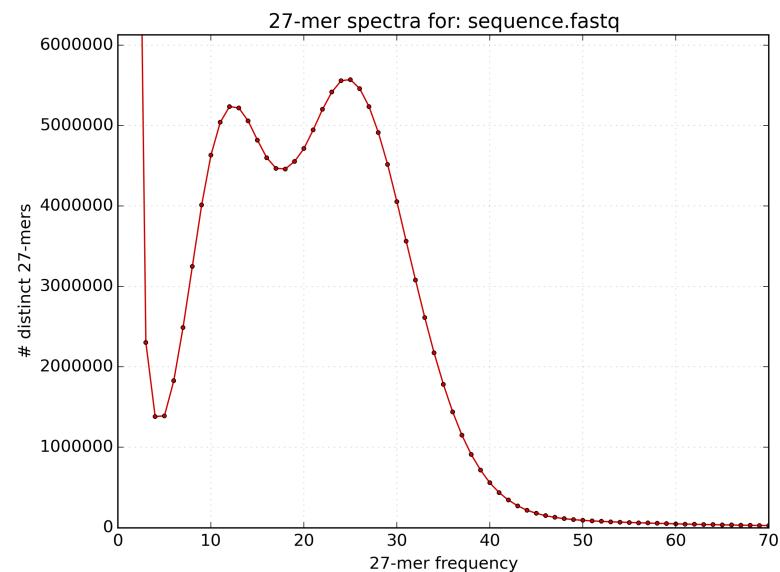
Frequency of those k-mers in  
the sequence data set



x-axis: Frequency of the  
distinct k-mer in the data  
y-axis: Count of distinct  
k-mers with frequency x

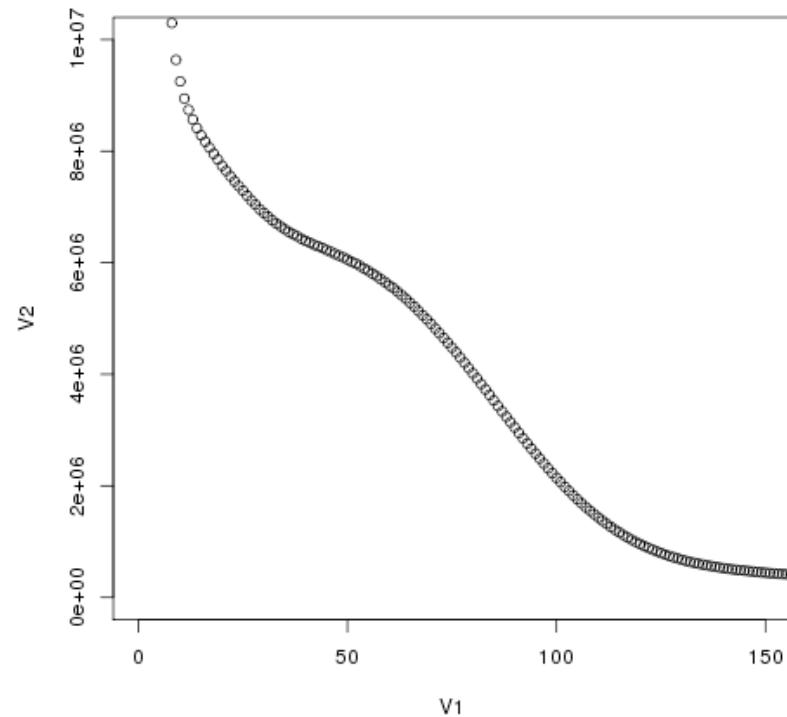
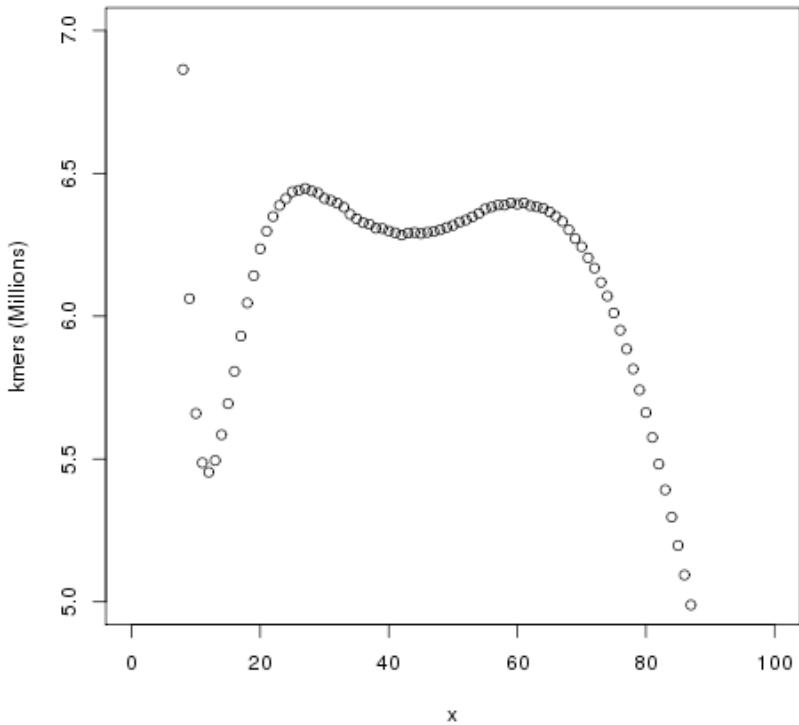
# Estimating Genome Size

- We can use coverage to estimate genome size
  - The peak with the largest k-mer multiplicity is the mean k-mer coverage across the genome.
  - $N = M * L / (L - K + 1)$ 
    - N is Depth of Read Coverage
    - M is mean k-mer coverage
    - L is read length
    - K is k-mer size
  - $G = T / N$ 
    - G is the genome size
    - T is the total number of bases



# Estimating Genome Size

- Not so easy: estimating complexity



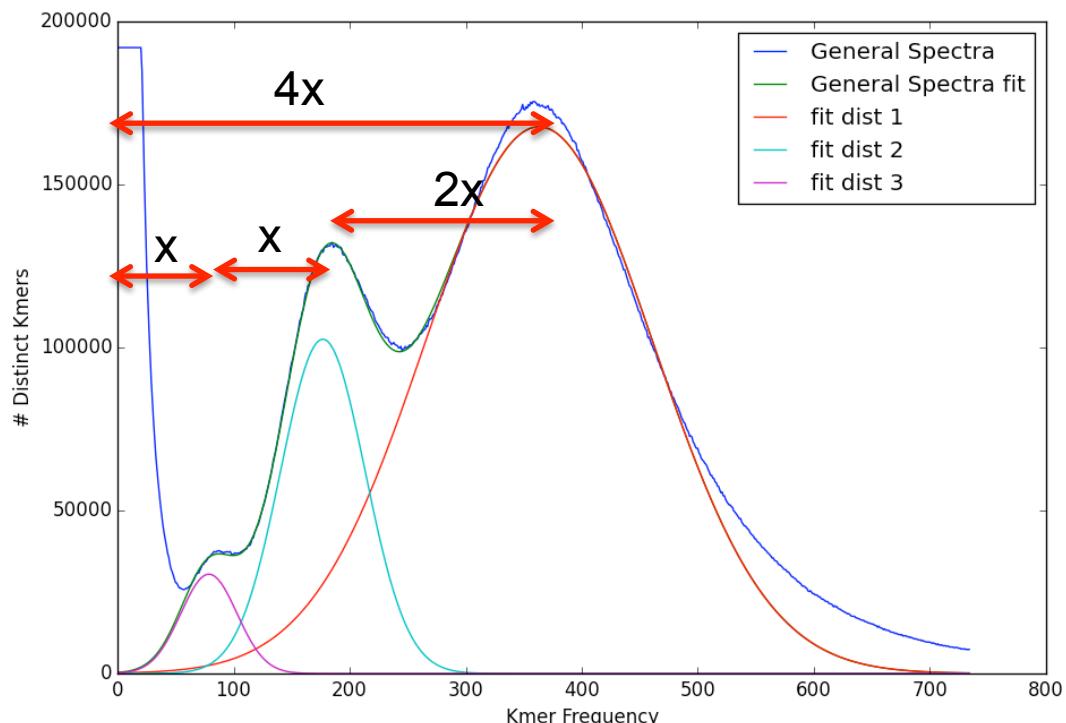
# Estimating Genome Size

- Distribution decomposition analysis
  - `kat_distanalysis.py --plot kat.hist`

```
Main spectra statistics
-----
K-value used: 27
Peaks in analysis: 3
Index Freq Max Volume
1 79 35596 1844622
2 177 132229 9354614
3 364 175454 41202694
Mean k-mer frequency: 320x
Homozygous peak index: 3
Estimated genome size: 48.05 Mbp
Estimated heterozygous rate: 0.86 %
Estimated assembly completeness: 95.73%
Breakdown of copy number composition for each peak
-----
--- Report for f=364.209 (total elements 36483749) ---
0x: No significant content
1x: 100.00% (36483749 elements at f=367.62)
2x: No significant content
3x: No significant content

--- Report for f=177.838 (total elements 9214569) ---
0x: 47.33% (4361617 elements at f=183.02)
1x: 52.67% (4852951 elements at f=176.43)
2x: No significant content
3x: No significant content

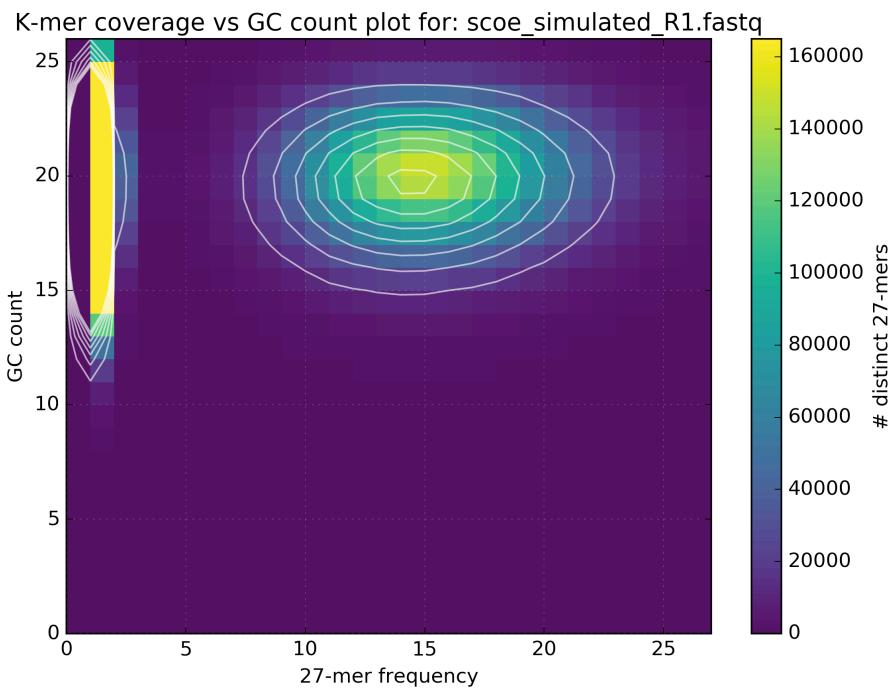
--- Report for f=79.618 (total elements 858966) ---
0x: 100.00% (858966 elements at f=77.18)
1x: No significant content
2x: No significant content
3x: No significant content
```



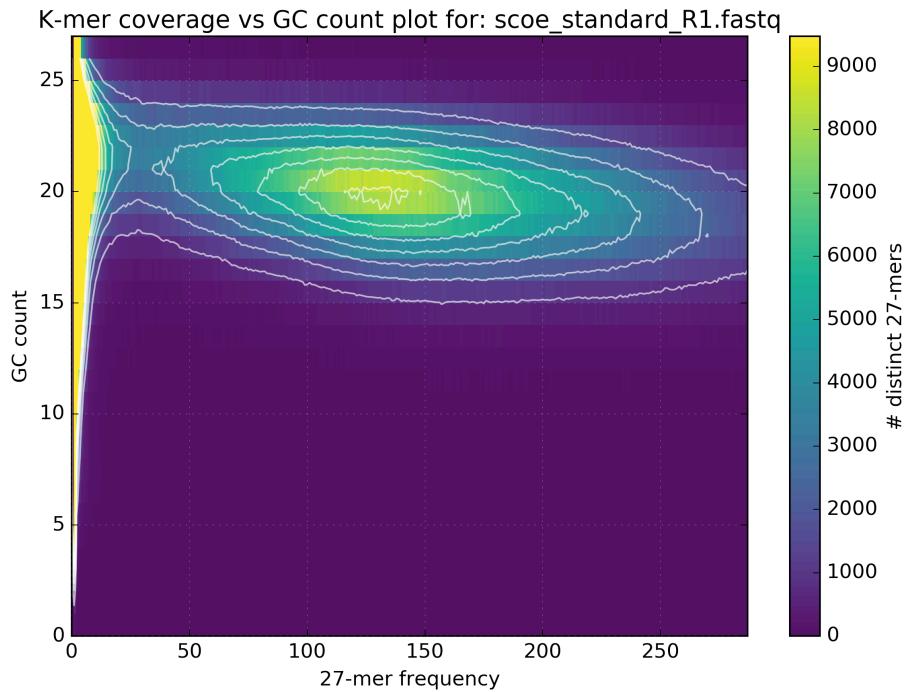
# GC Content in depth

- Monitor GC Bias

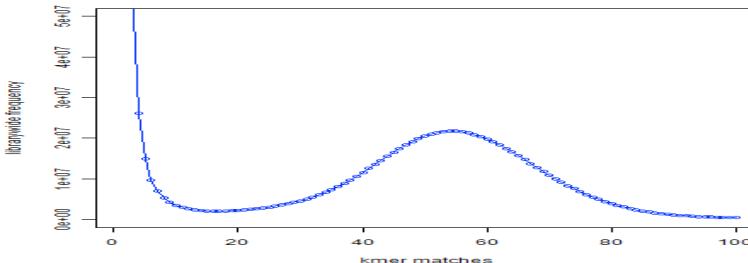
## Simulated data



## GC bias in Standard Illumina protocol



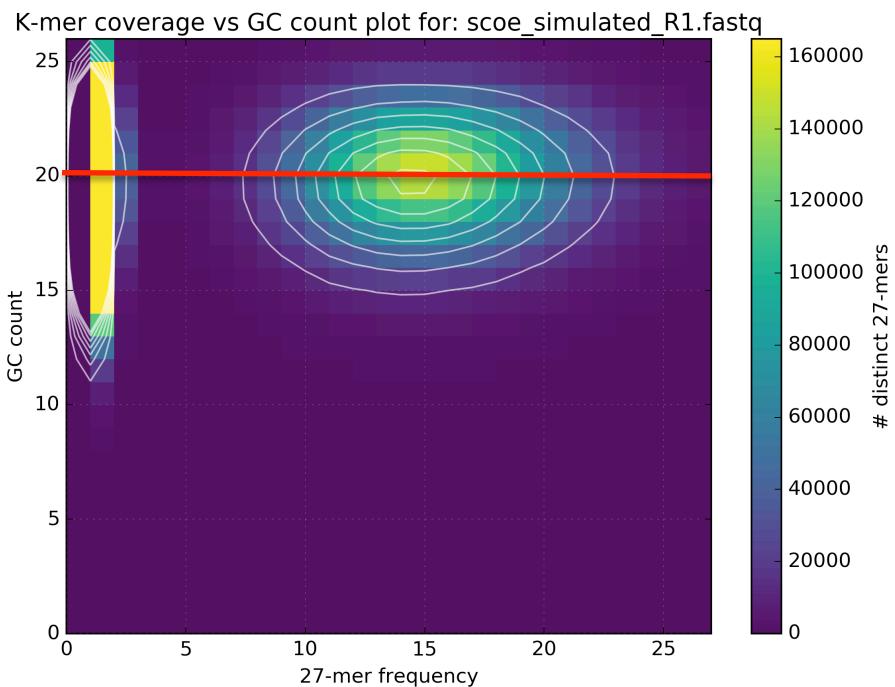
ACTCAGGGATTA    GC=4 Count=1  
AATAGCCGGGG    GC=7 Count=2



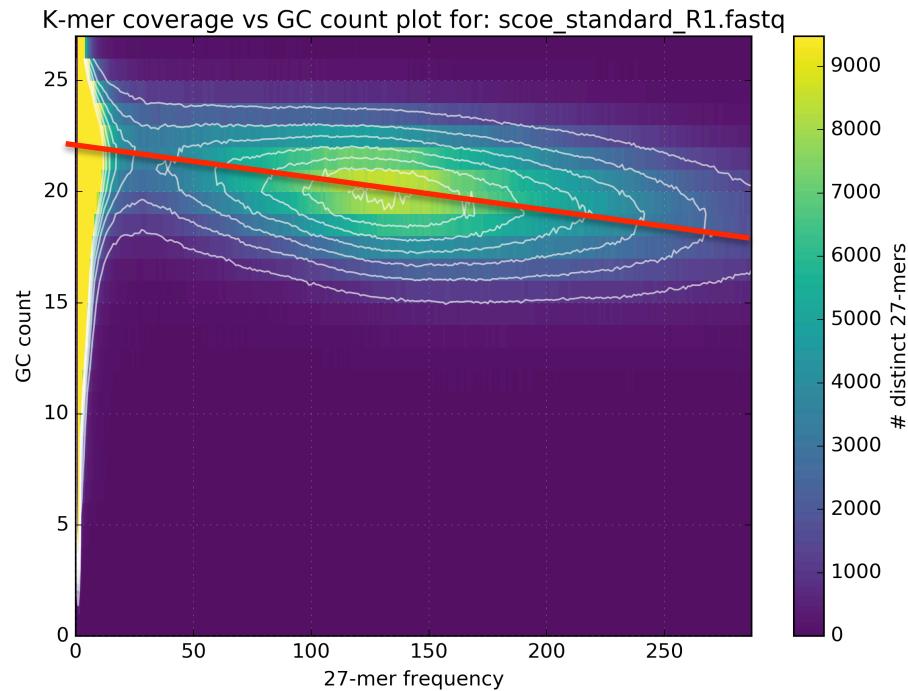
# GC Content in depth

- Monitor GC Bias

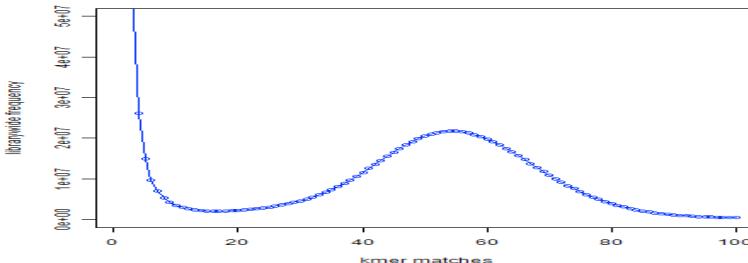
## Simulated data



## GC bias in Standard Illumina protocol

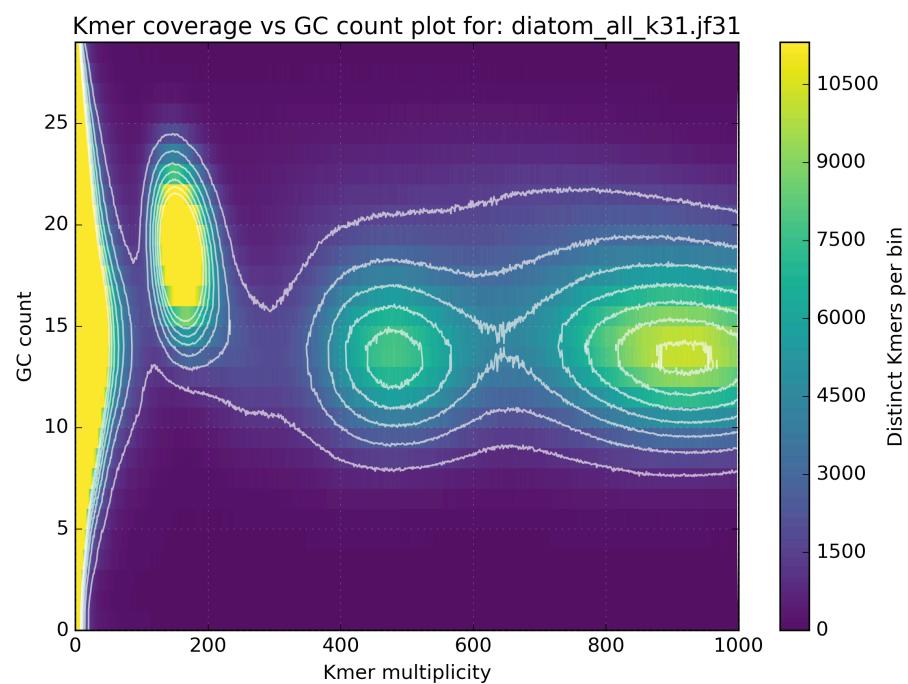
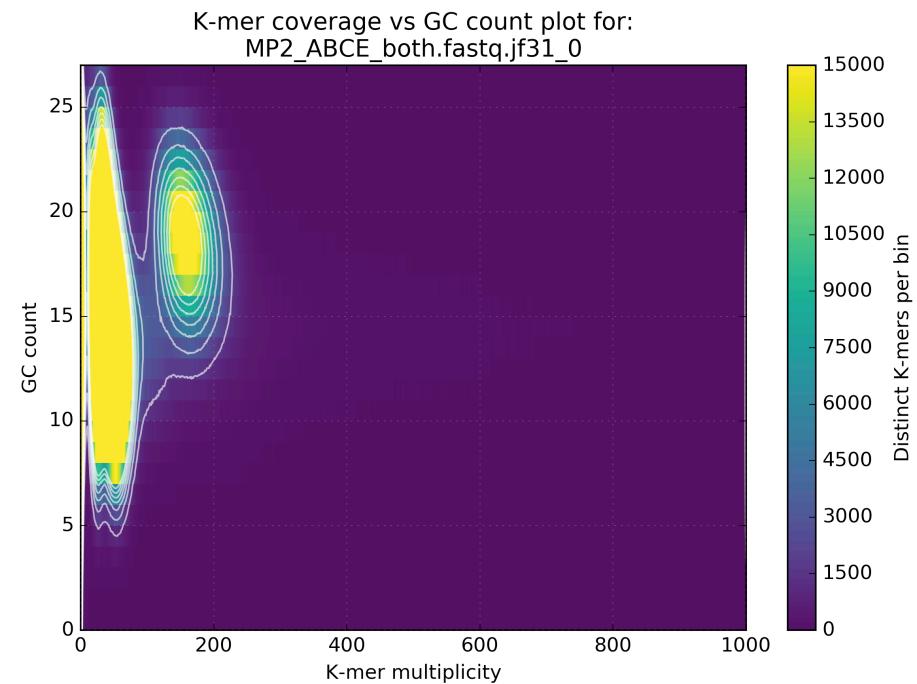


ACTCAGGGATTA    GC=4 Count=1  
AATAGCCGGGG    GC=7 Count=2



# GC Content in depth

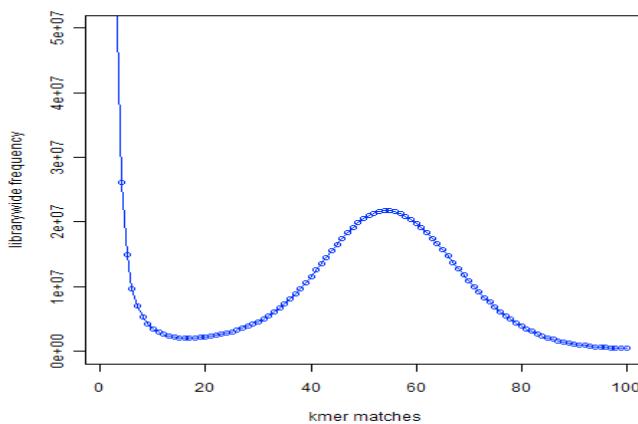
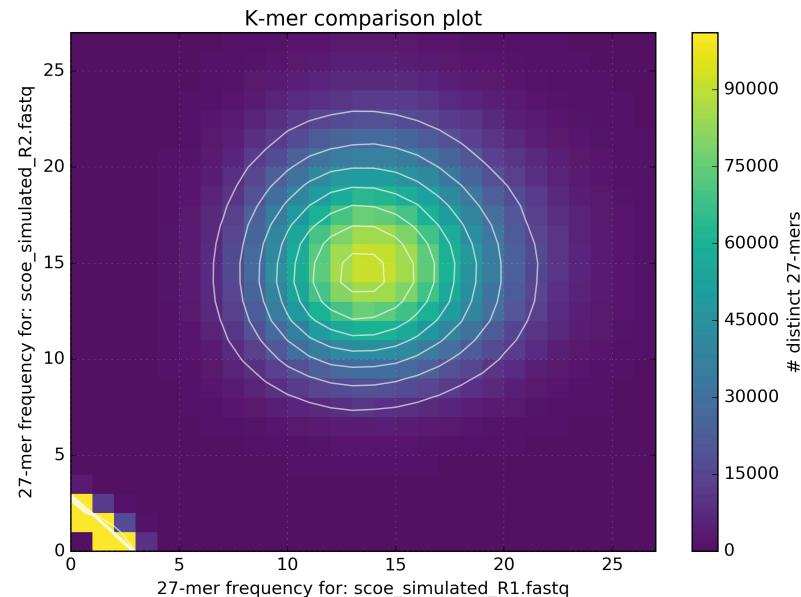
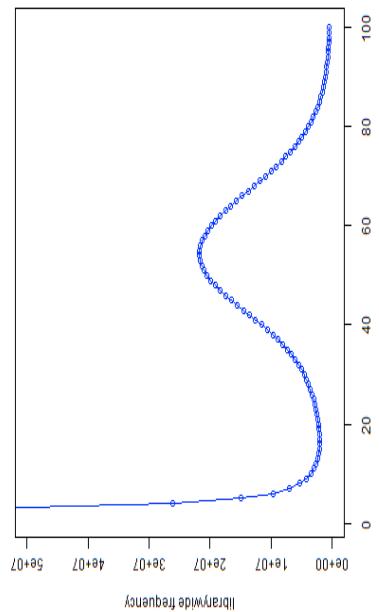
- Uncover contamination
  - Separate bacteria from eukaryote
  - Separate organelle from nuclear genome



# What can k-mers tell us?

- Comparing k-mer counts between data reveals biases

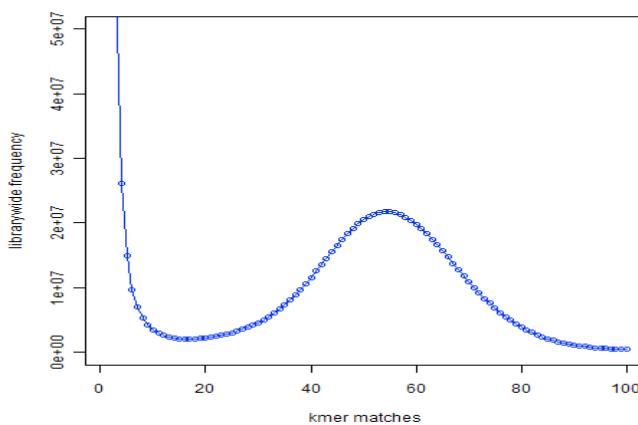
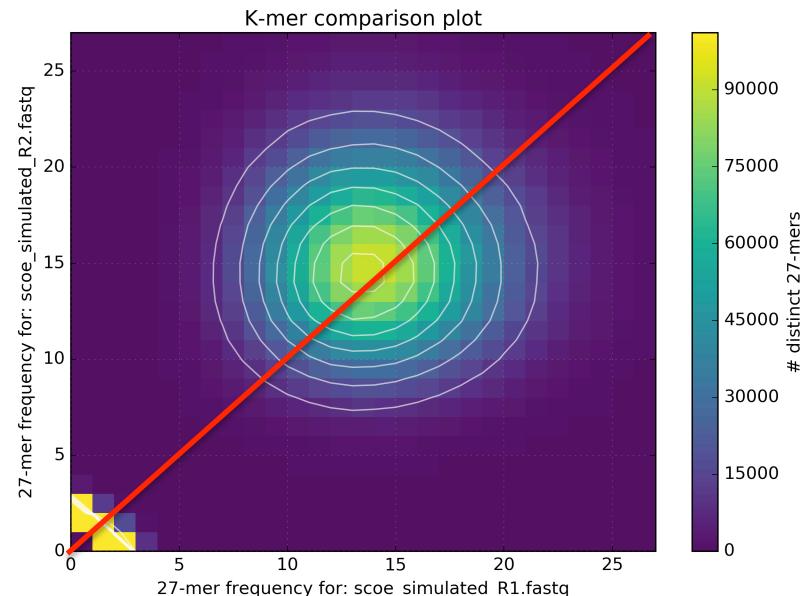
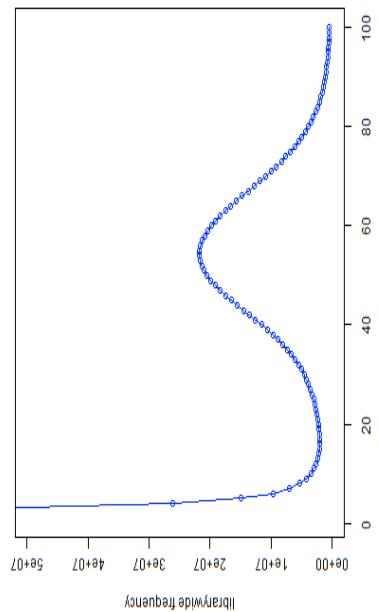
- R1 vs R2
- Lib1 vs Lib2



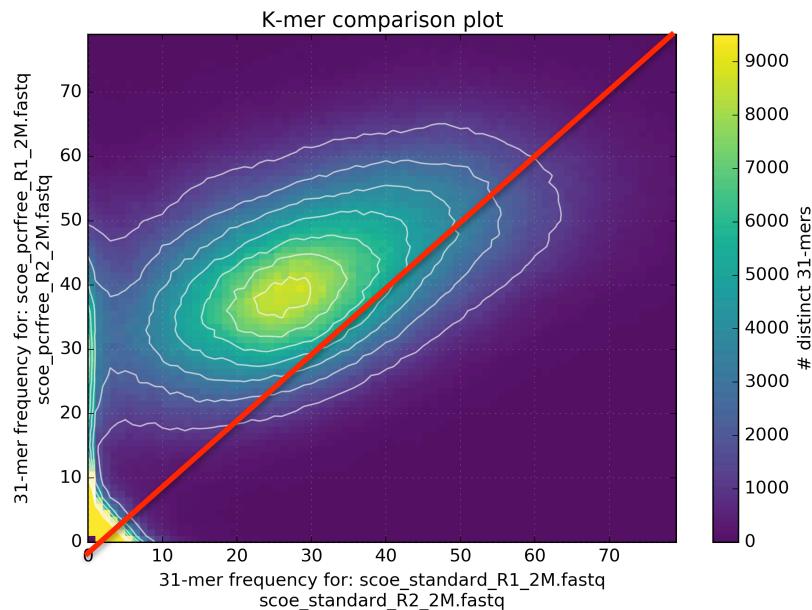
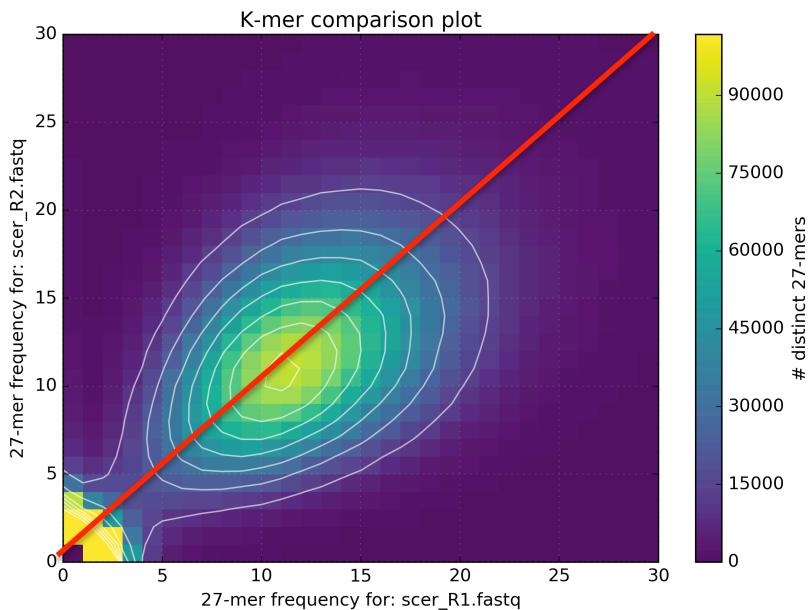
# What can k-mers tell us?

- Comparing k-mer counts between data reveals biases

- R1 vs R2
- Lib1 vs Lib2



# Data comparison

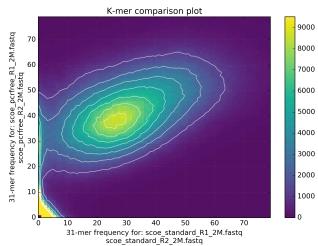
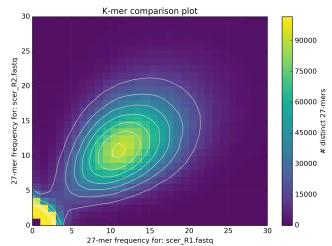
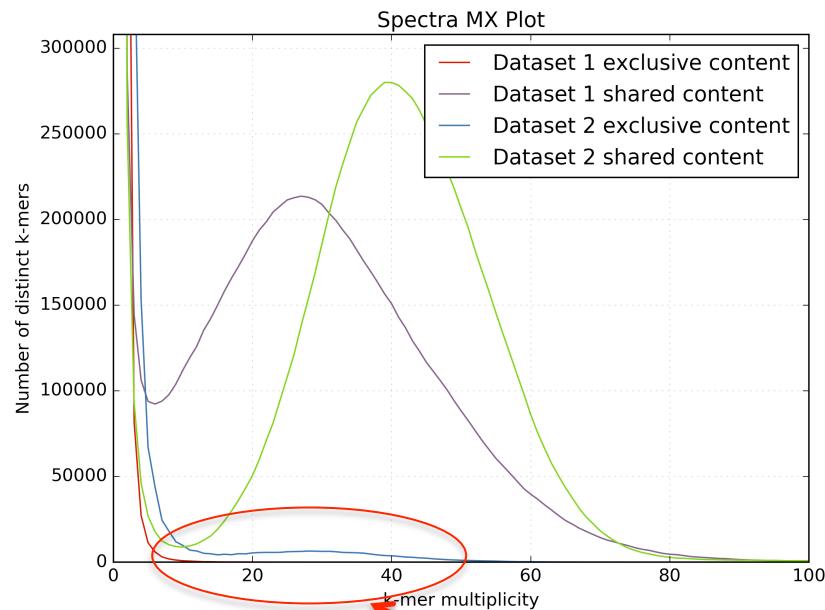
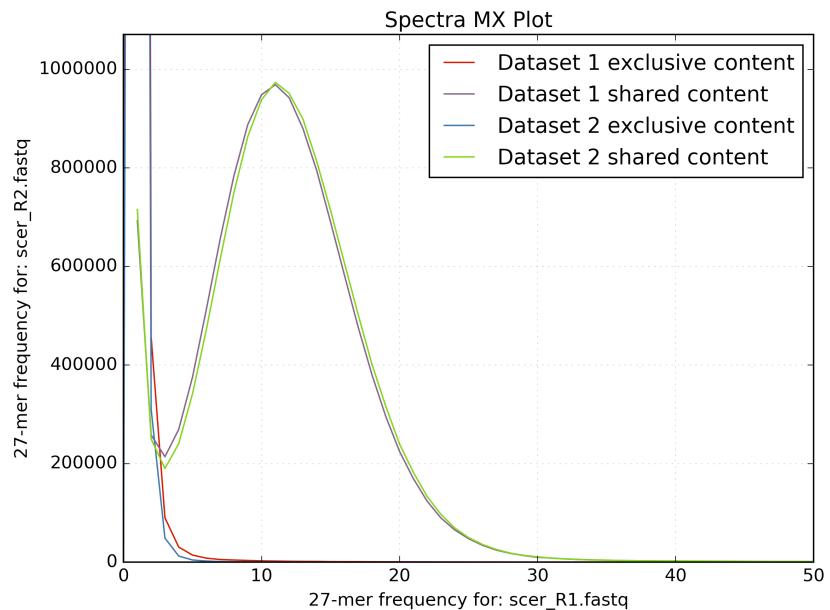


R1 vs R2

Standard vs PCR free

- PCR free captures data missing in standard protocol

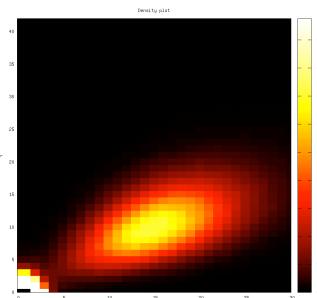
# Data comparison



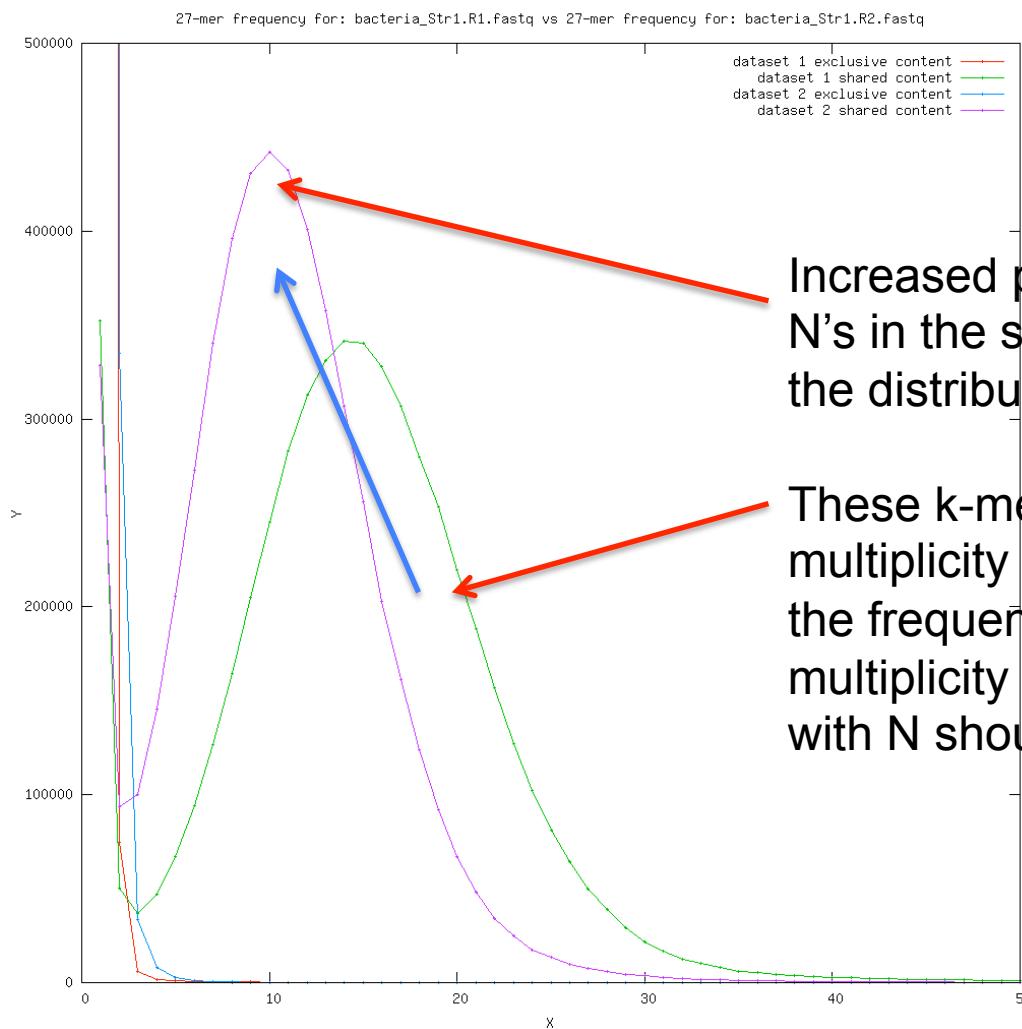
Low proportion of k-mers present only in dataset 2

# Data comparison

R1 v R2



Lower quality dataset =  
More N's in  
the sequence

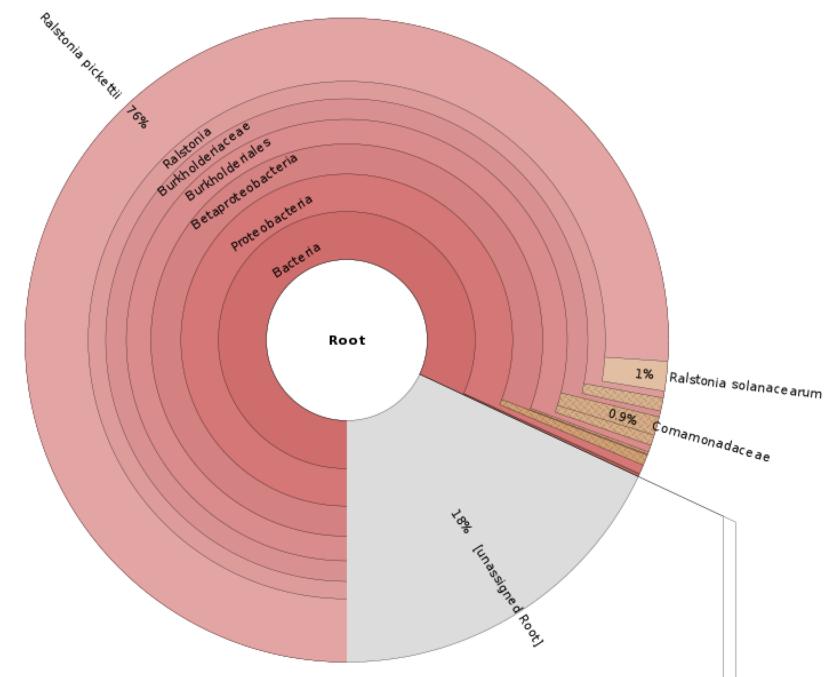
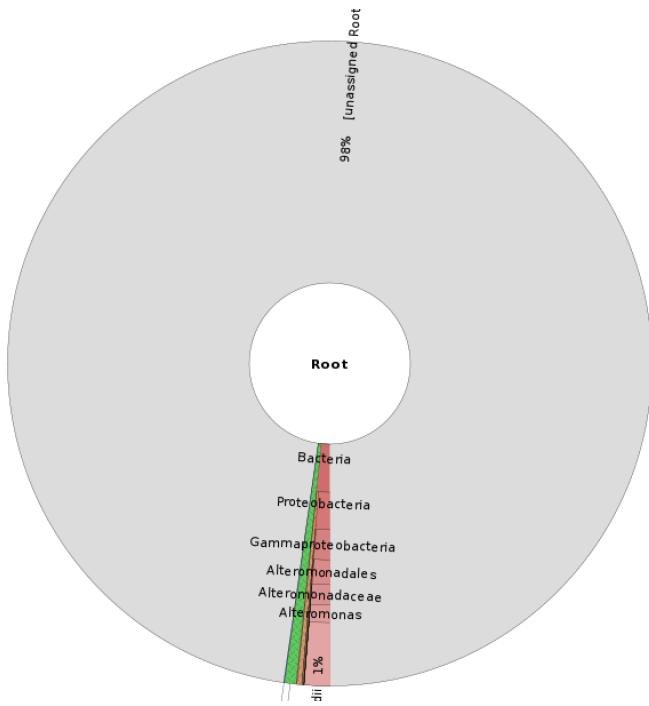


- K-mer based analyses provides a lot of information about the information within your data.
- K-mer filtering can be used to assist exploratory analyses.
  - Digital normalization can help to process uneven coverage.
  - Error correction can be used to remove low frequency errors.
  - Frequency based filtering can separate organelles or contamination for assembly.

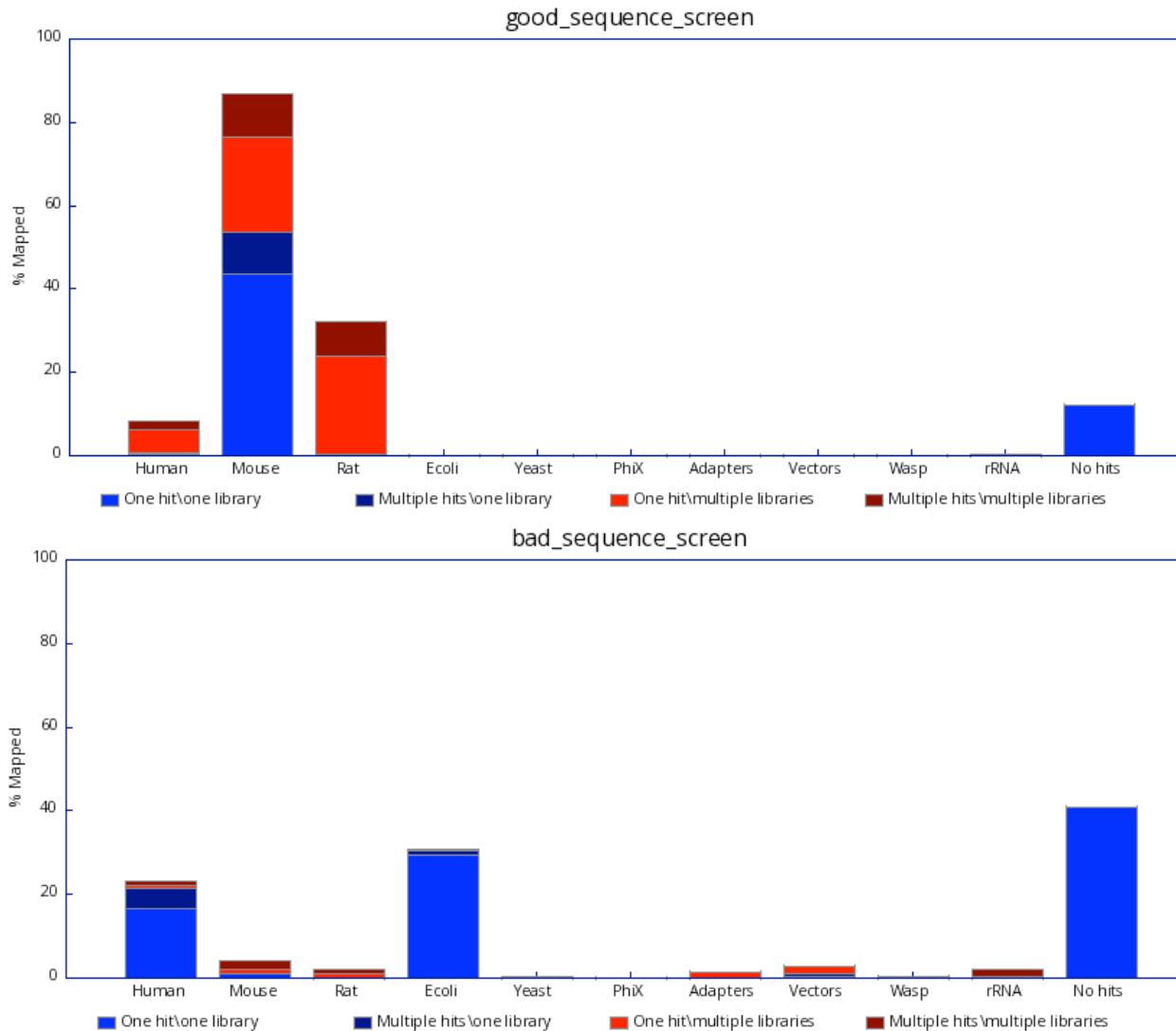
- PacBio and Nanopore have greater error rates per base. This precludes many analyses that one can do with Illumina data.
  - K-mer analyses are not feasible because there are too many unique k-mers in the reads.
    - Reads of Insert / CCS reads require at least 30x coverage in order to increase the accuracy to the necessary threshold.
  - Biases and data differences are more difficult to detect.
  - K-mer based contamination analyses have lower accuracy.

# Contamination Analyses

- Read based contamination analyses are tricky
  - Entirely dependent on your reference database
  - Short k-mer matching increases alignment to multiple targets
  - Unrelated organisms can contain similar strings of nucleotides



# FastQ Screen



- K-mer analyses are limited by:
  - sufficient depth of coverage
  - sequence error rates
- K-mer analyses can:
  - Help estimate genome size and infer ploidy
  - Detect library biases within and between data sets
  - Help find contaminants
- K-mer based filtering can make data easier to work with
- Contamination assessment:
  - Entirely dependent on your subject database
  - Loss of accuracy with shorter strings