Ubiquitous Genomics: Hackathon 1

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Problem 1

Number of 1D and 2D reads classified as failed: 3364 Number of 1D and 2D reads classified as passed: 3243

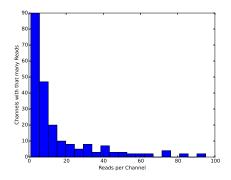
Number of 2D reads classified as failed: 540 Number of 2D reads classified as passed: 1081

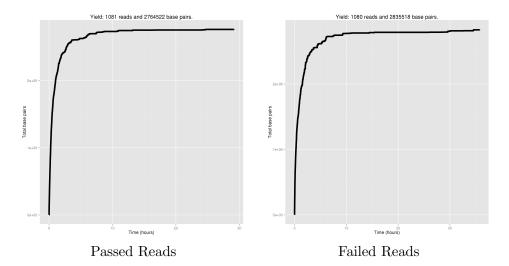
Fraction of 2D reads in failed folder: 0.160523186683 Fraction of 2D reads in passed folder: 0.333333333333

Problem 2

218 channels had at least one read. The average channel had 14.8 reads. Channel 212 had 95 reads, which was the most.

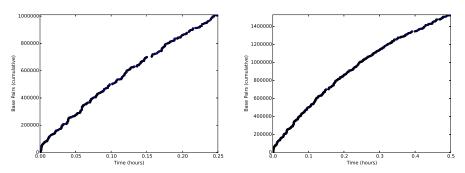
Just for fun, here's a histogram of reads per channel





Problem 4

Let us zoom in a little on the previous graphes:



The first quarter hour appears linear, but the first half hour does not. Over the first half hour, the rate was 1121.4 base pairs per second. Therefore, to handle the 3 billion base pairs of the human genome would take 743.1 hours.

Passed 2D Reads

Mean Quality Score=10.3775 Std. Dev.=1.83151 n=2764522

Failed 2D Reads

Mean Quality Score=9.48623 Std. Dev.=2.06174 n=1417759

T-test comparison

T-Statistic=451.124 P-Value=0

The P-value was confirmed to be returned as 0 from Python/numpy. We believe that the true value is above 0 but beneath the lowest threshold of Python's float value (2.2250738585072014e-308).

Passed 2D Reads - First Hour

Mean Quality Score=10.3668 Std. Dev.=1.82208 n=1970130 Median=10

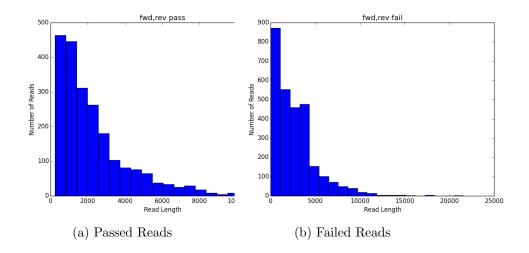
Passed 2D Reads - Last Hour

Mean Quality Score=11.5651 Std. Dev.=2.12032 n=968 Median=11

The quality of the reads does not appear to have degraded over the timespan of the sequencing (on the contrary, there is a slight increase).

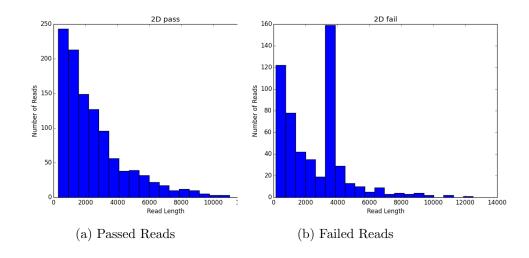
1D reads

The following histograms show the length distribution of 1D reads (both template and complement) for passes and fails.



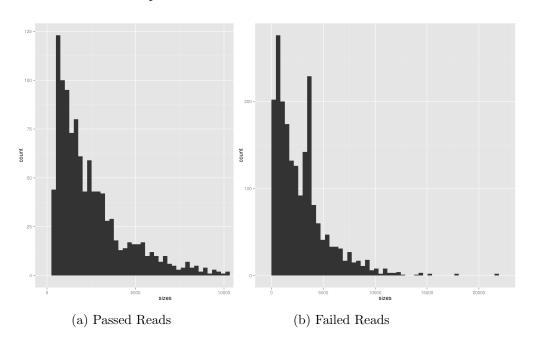
2D reads

The following histograms show the length distribution of 2D reads for passes and fails.



1D and 2D reads

The following histogramss show the cumulative length distribution of both 1D and 2D reads for passes and fails.



Problem 7

LONGEST TEMPLATE READ

From file: MINION02_4teamawesome_2446_1_ch312_file44_strand.fast5

Number of nucleotides: 11820

LONGEST COMPLEMENT READ

 $From file: MINION02_4 team awe some_2446_1_ch312_file44_strand.fast5$

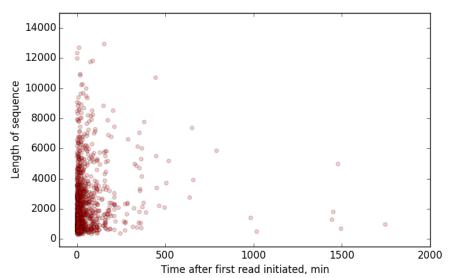
Number of nucleotides: 11498

LONGEST 2D READ

 $From file: MINION02_4 team awe some_2446_1_ch312_file44_strand.fast5$

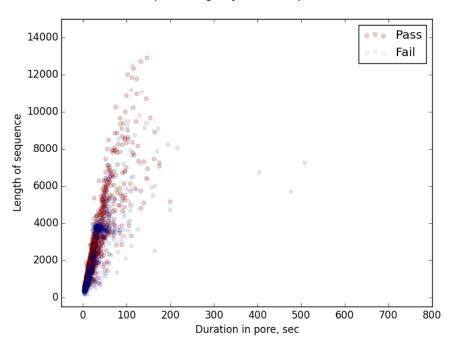
Number of nucleotides: 12916

Sequence length by time entered pore



☐This graph was generated from the 2D pass reads, and time was ☐calculated as time (in minutes) that each read was initiated ☐from the time the first read was initiated. Those reads that ☐initiated late did not achieve a long read length.

Sequence length by duration in pore, sec



Problem 10

Computing nucleotide composition of passed reads...

Failed reads %Composition

% A: 26.8544936058 % C: 22.6248607838 % T: 26.8972371186 % G: 23.6234084919

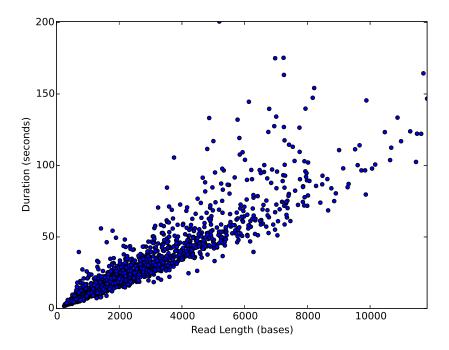
Computing nucleotide composition of failed reads...

Passed reads %Composition

% A: 27.8932849874 % C: 21.7440845108 % T: 27.990263778 % G: 22.3723667238

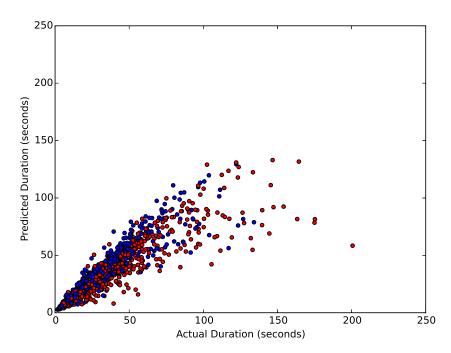
Simple Model

The simplest possible model is the one in which the number of nucleotides determines the read duration, possibly with some constant term for getting started. To consider this, we start with a scatterplot of bases against time with a linear fit.



We might be tempted to include a constant term in our fitting, but as should be apparent, it would be negative. How long it would take to sequence an extremely short read is unclear. Fortunately, there are none in our sample. This allows us to construct a trivial (and still pretty effective) model based on a fixed cost per nucleotide.

Cost per nucleotide: 11.25ms

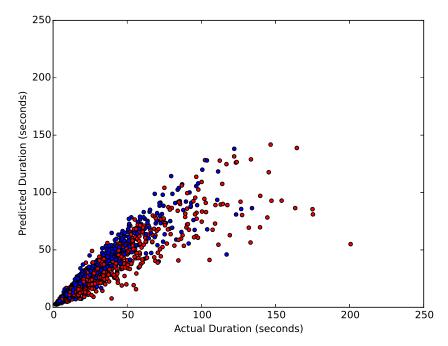


 $r^2 = 0.84$

(Red circles are template strand; blue are complement. There appears to be no significant difference between them.)

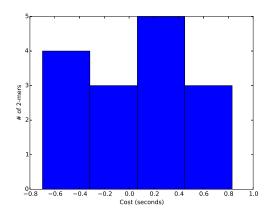
Nucleotide Model

A slightly more complex model uses a differend cost for each type of nucleotide: A=32ms C=-07ms G=04ms T=12ms

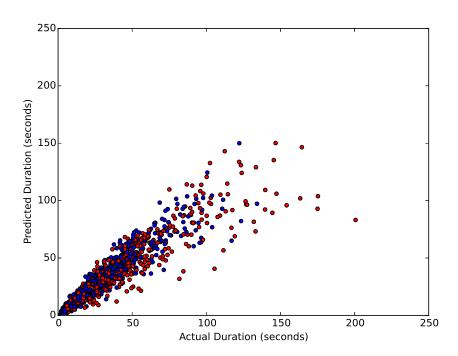


 $r^2 = 0.84$

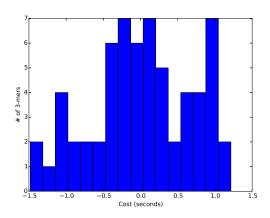
We can use a model in which each the time to extend by one nucleotide is determined by the 2-mer in the middle of the pore. We can not list all the costs, but here's a histogram:



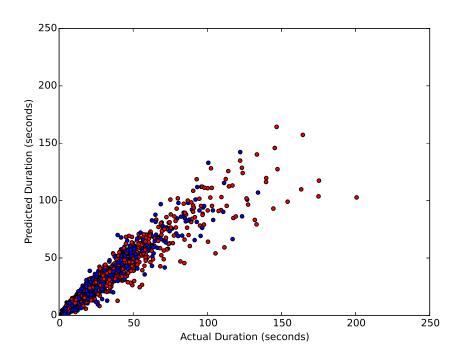
And here's the resulting predictions:



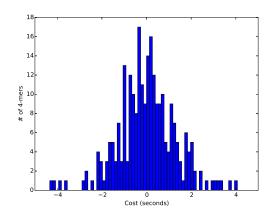
We can use a model in which each the time to extend by one nucleotide is determined by the 3-mer in the middle of the pore. We can not list all the costs, but here's a histogram:



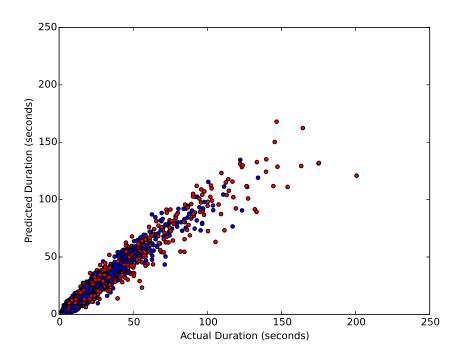
And here's the resulting predictions:



We can use a model in which each the time to extend by one nucleotide is determined by the 4-mer in the middle of the pore. We can not list all the costs, but here's a histogram:

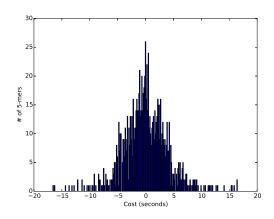


And here's the resulting predictions:

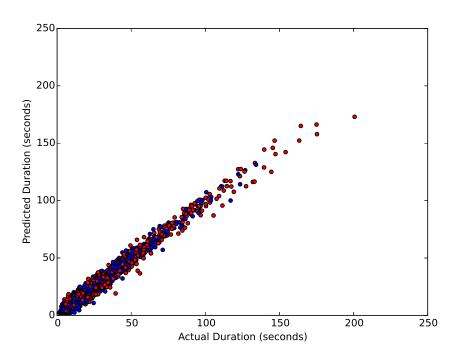


$$r^2 = 0.94$$

We can use a model in which each the time to extend by one nucleotide is determined by the 5-mer in the middle of the pore. We can not list all the costs, but here's a histogram:



And here's the resulting predictions:



 $r^2 = 0.98$

At this point our model has 1024 degrees of freedom for 2162 datapoints, so a good fit may reveal more overfitting than model appropriateness.