

Base Calling: methods, problems and alternatives

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EMBL Advanced Course in Analysis of Short Read Sequencing Data
8th June 2009 -- 10th June 2009

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Outline

New sequencing technologies

Image analysis

Lasers and cross talk

Chemistry and phasing

AYB -- a new base calling algorithm

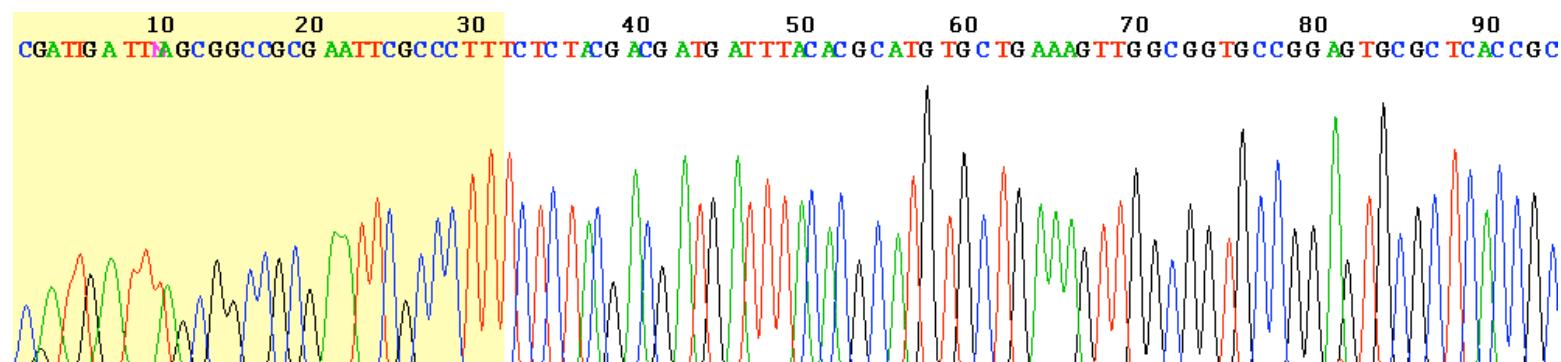
Quality calibration and sequencing errors

AGATAGGAAGAGCGGTTCAGGAAATGCCGAGA
Capillary sequencing

AB 3730xl Cutting edge of capillary technology
96 capillaries in parallel



Rapid	2100 kb/day	550 base reads
Accurate	690 kb/day	900 base reads



Images: <http://www.appliedbiosystems.com/>
http://en.wikipedia.org/wiki/File:Sanger_sequencing_read_display.gif

AGATAGGAAGAGCGGTCACCCACGGATGCCGAGA
454 Life Sciences (Roche)



454 GS FLX titanium

Current performance
1 million reads per run
400bp meanlength, at Q20

Future (454 lab demonstration)
650bp mean length (750 modal)

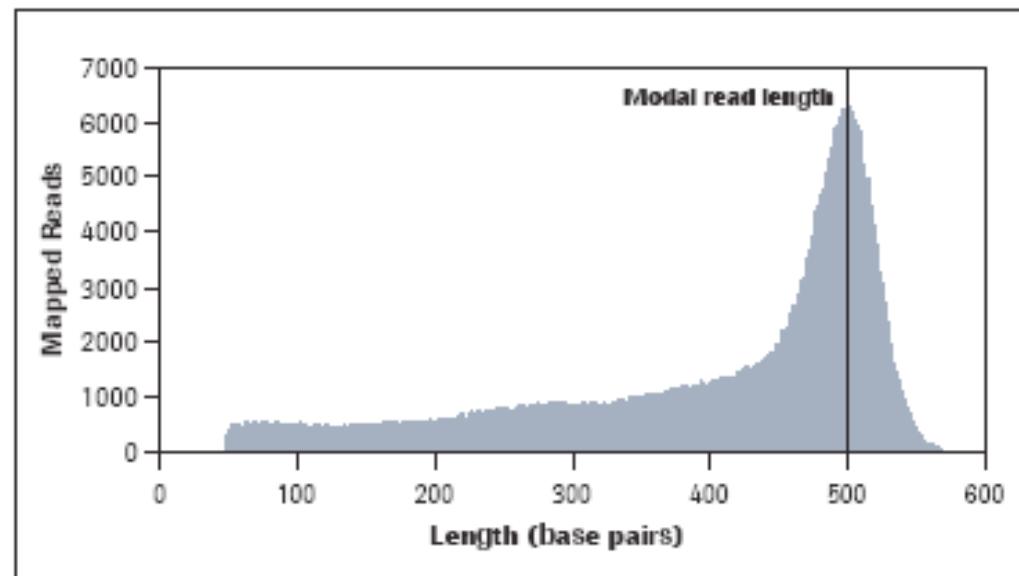


Figure 1: Example Read Length Distribution of 629,643 reads from *E. coli* K-12 (Genome size ~4.5 Mb) with a modal read length of 504 bases.

Images: <http://www.454.com/>

AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA
Illumina GA II



Current performance

14 Gb per run 2 x 75bp

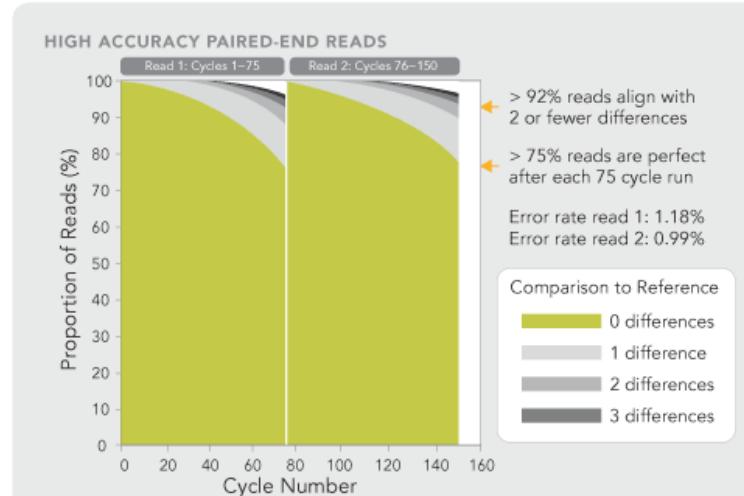
20 Gb per run GA II_x software upgrade

150bp reads demonstrated outside Illumina

Future (2010) with new array technology

"sub-micro semi-ordered array"

55 Gb -- 100 Gb 2 x 125bp,



The Genome Analyzer provides a powerful combination of high output quantity and quality. This graph depicts the high per base accuracy profile from a recent 14.1Gb run with 2 x 75bp paired-end sequencing. Both reads show equivalently high rates of perfect reads (> 75%) and reads with two or fewer differences (> 92%). Results were internally generated using the current Genome Analyzer_{II} System.

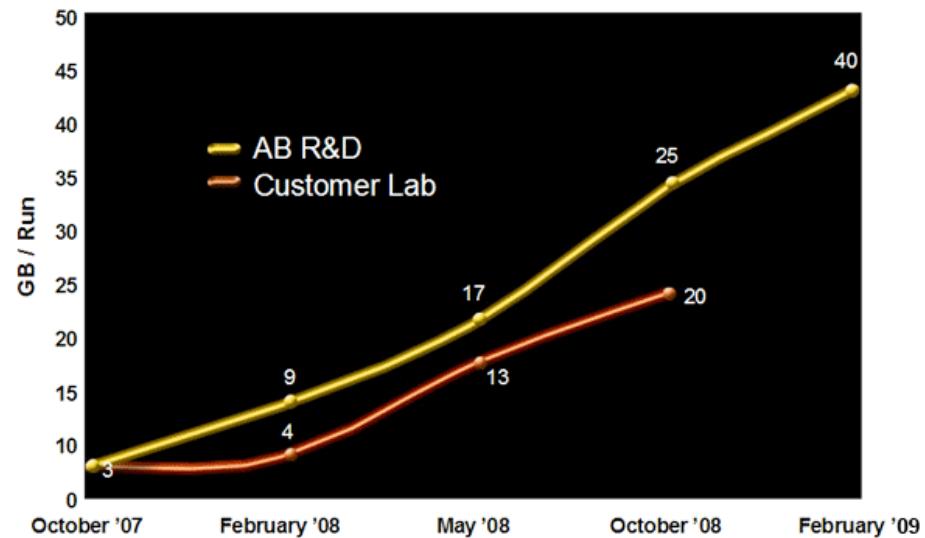
Images from <http://www.illumina.com/>

AGATAGGAAGAGCGGTCACCAAGAATGCCGAGA
Life Technologies' SOLiD



20-30 Gb 2 x 50bp (~ 2 weeks)

Industry-Leading Throughput



AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA
Helicos

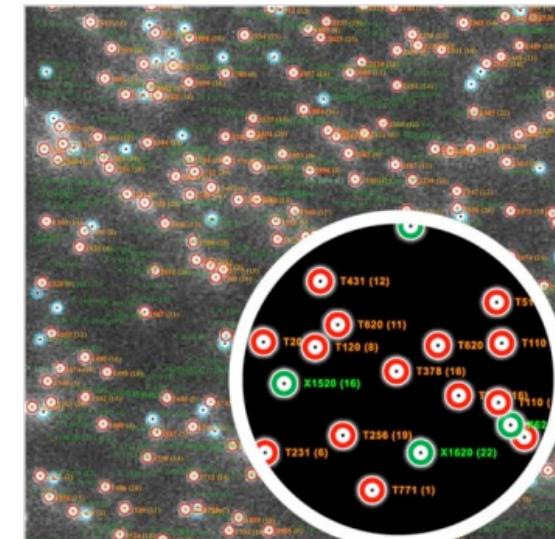
Single molecule sequencing -- no amplification



21-28 Gb per run, 105-140 Mb per hour
Read length 20 to 55bp, 30-35bp average

Asynchronous: separate steps for A, C, G, T

- strands get sequenced at different rates
- base composition bias in length of read

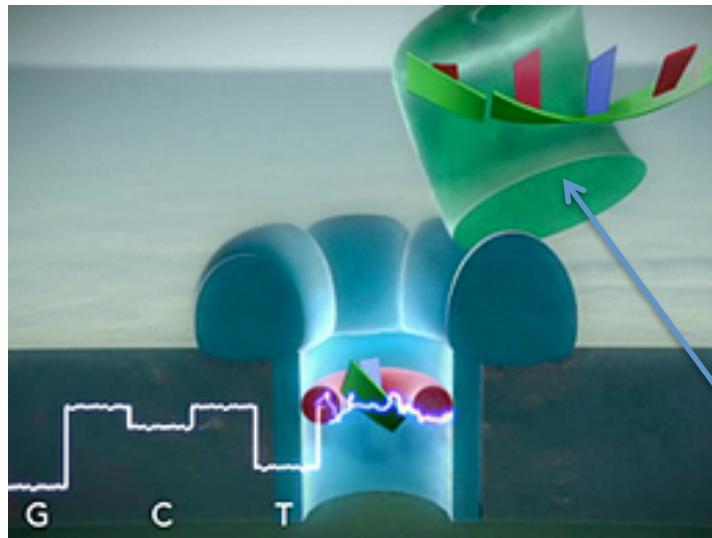


Images: <http://www.helicosbio.com/>

AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

Coming technology

Oxford Nanopore



Nanopore sequencing

No fluorophores

- use electrical properties of base passing through pore
- Can detect methyl-cytosine

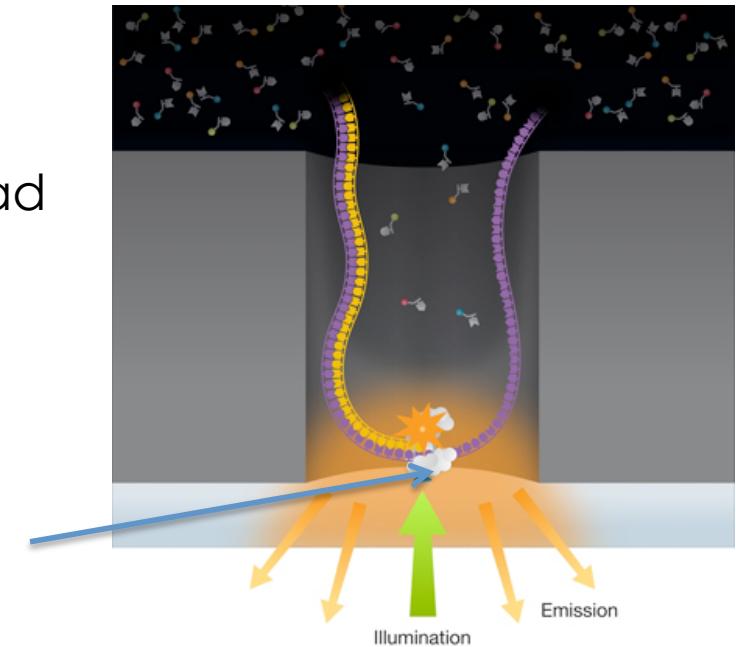
Aims

50bp/sec per read
Kb length reads

100 Gb per hour

modified
polymerase

Pacific Biosciences



Zero-mode wave guide

Tiny illuminated volume

- only bound fluorophores contribute
- watch incorporation in real-time, including errors

Images: <http://www.nanoporetech.com/>
<http://www.pacificbiosciences.com/>

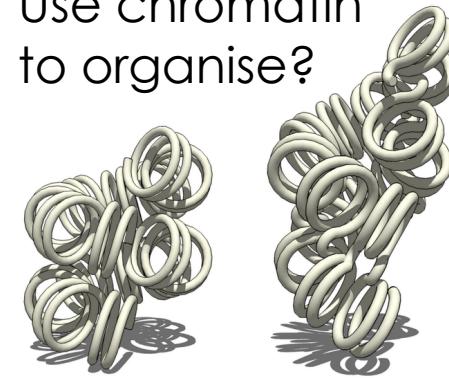
AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA
Limits on read length



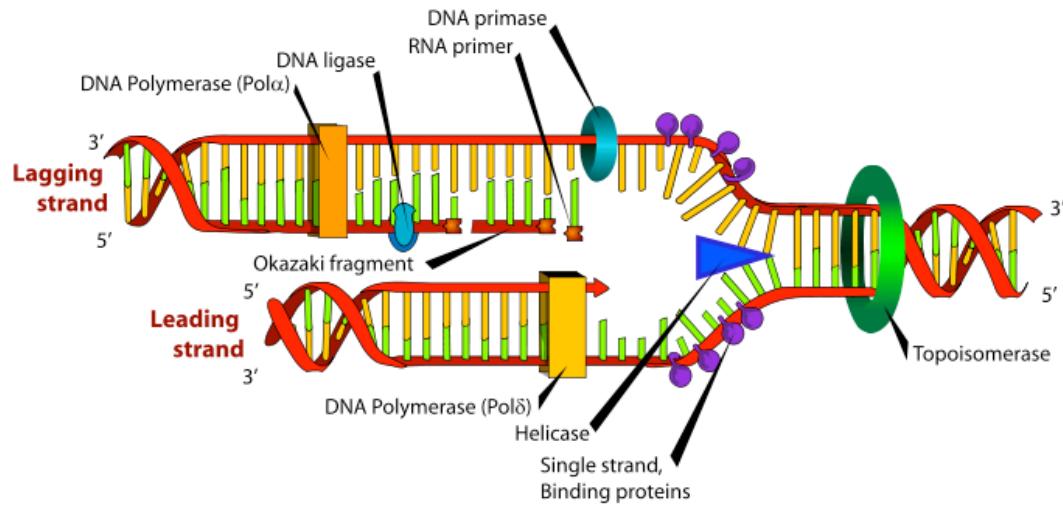
DNA spaghetti

- knots
- snaps if tugged
- sticks to walls when cooked

Use chromatin
to organise?



DNA replication



Images

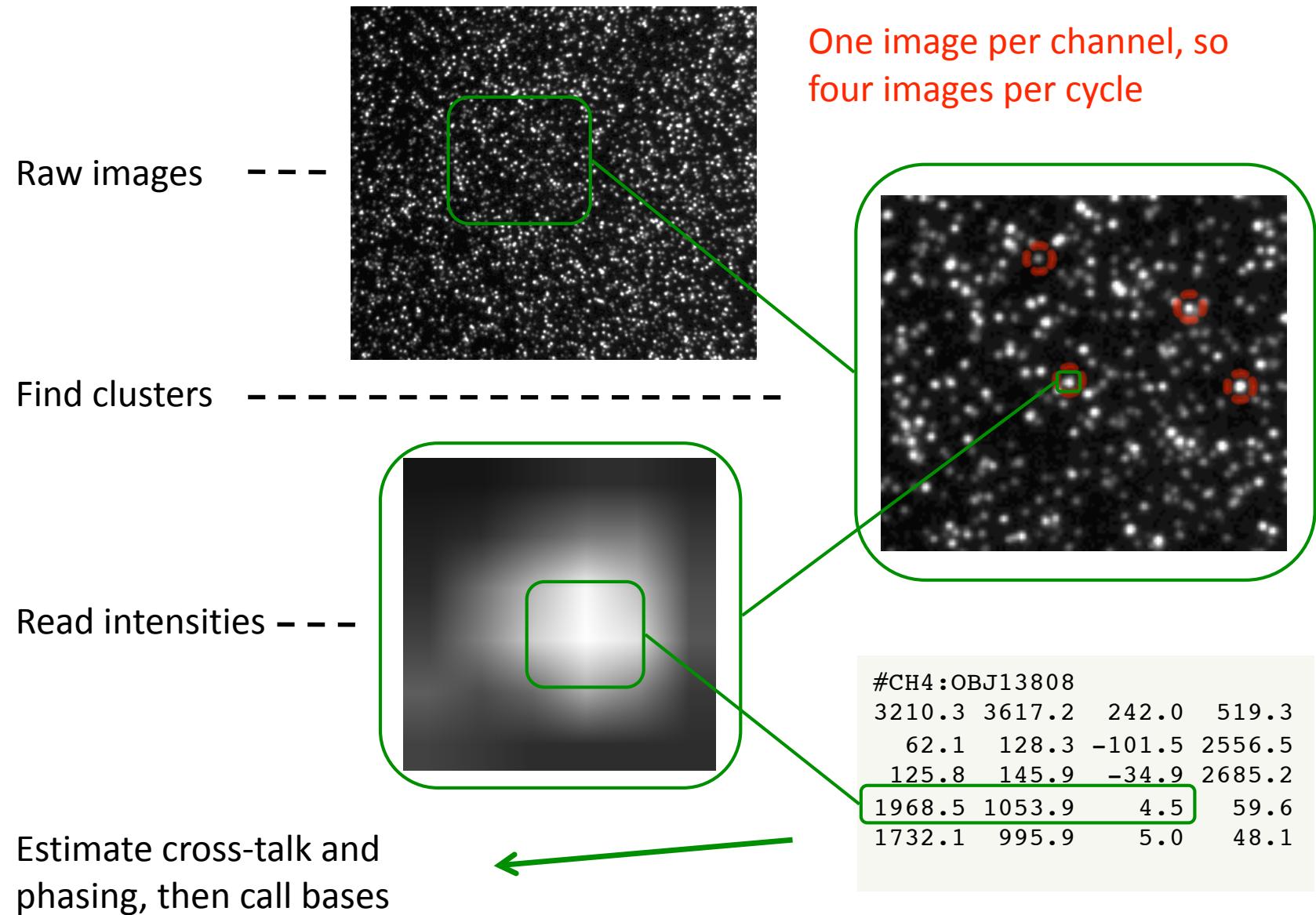
<http://en.wikipedia.org/wiki/File:Spaghetti.jpg>

http://upload.wikimedia.org/wikipedia/commons/6/6a/30nm_Chromatin_Structures.png

http://en.wikipedia.org/wiki/File:DNA_replication.svg

AGATAGGAAGAGCGGTTCA
GGCAGGAATGCCGAGA

Analysis pipeline



AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Image analysis

Registration

Filtering

- Sharpen clusters
- Edge detection

Normalization

- subtract background
- noise

Cluster identification

- deblend (split large clusters)
- remove local background

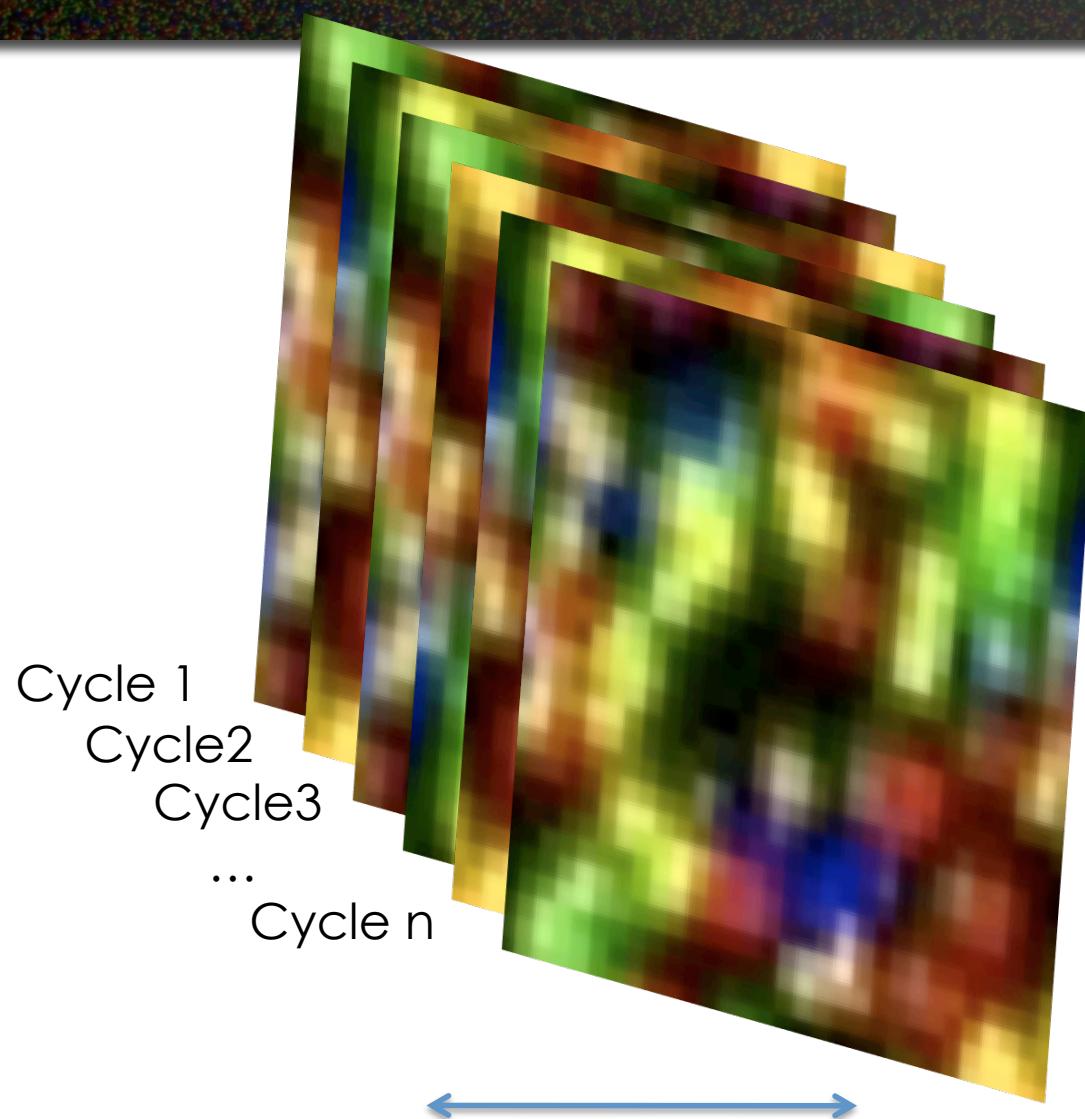
Warning: out of date, describes older version of pipeline

Based on notes prepared by Nava Whiteford,

<http://sgenomics.org/mediawiki/upload/8/80/Pipeline.pdf>

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Image Registration



Clusters “move” between cycles

- instrument jitter
- focal changes

Must track clusters between cycles and align images

AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

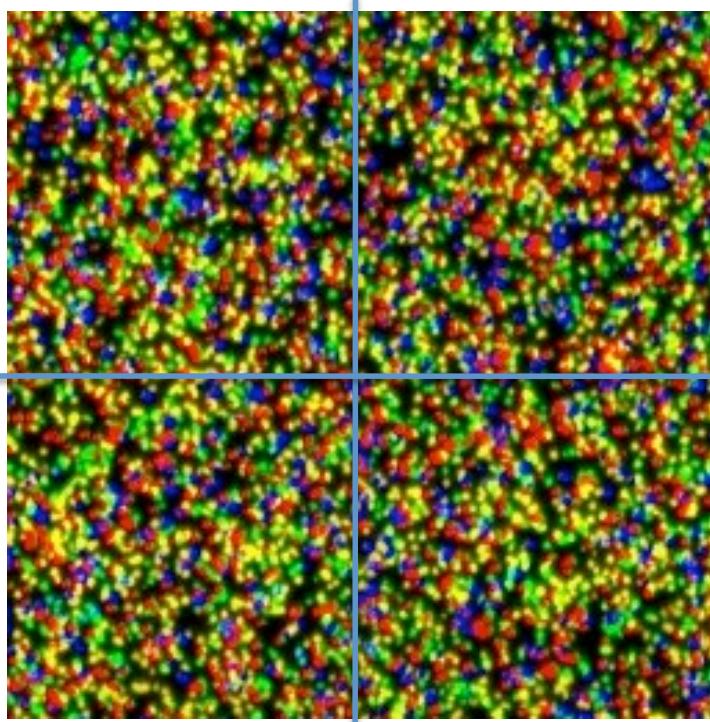
Image Registration

Three effects to compensate for

- Translation
- Rotation
- Scaling

}

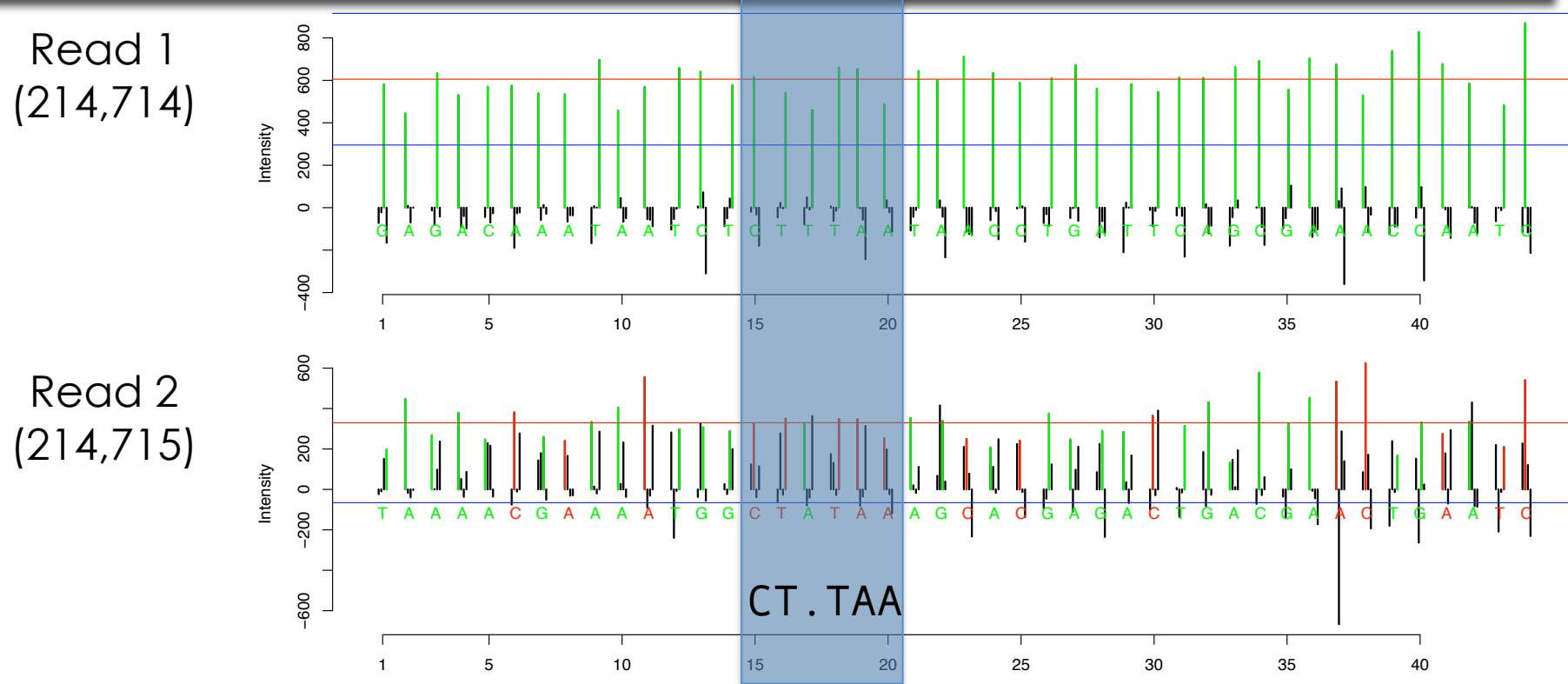
Restricted form of affine transformation
(procrustes transformation)



Basically a dynamic programming problem

Split image into regions
Estimate transformation for each region
Take consensus

AGATAGGAAGAGGCCGTTCAGCAGGAATGCCGAGA



Do errors in read match their neighbour? p-value ~ 1e-4 (0.83)

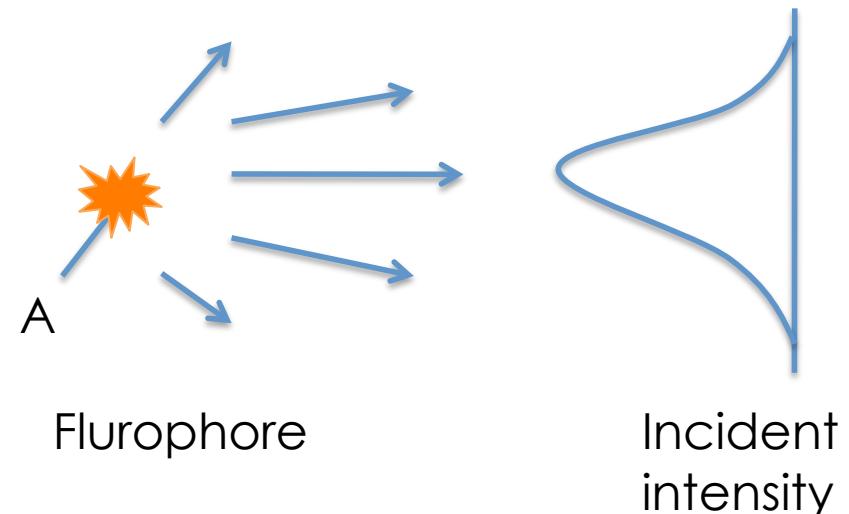
	Match	Mismatch
Observed	12 (8)	4 (20)
Expected	4 (7)	12 (21)

(Corresponding numbers for positions correct in read 2 are in brackets)

AGATAGGAAGAGCGGTTCA GCACGGAAATGCCGAGA

Filtering

- Clusters become blurred
- emitted light not entirely coherent
 - focal problems



- Need to correct for blurring to find position of cluster and emitted light
- sharpening
 - edge detection



Original cluster

Gaussian blur

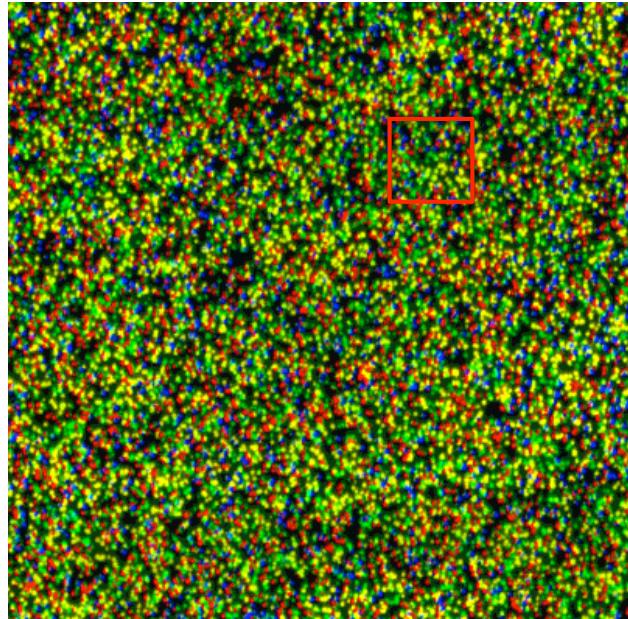
AGATAGGAAGAGCGGTTCA GCACGGAAATGCCGAGA

Convolution filters

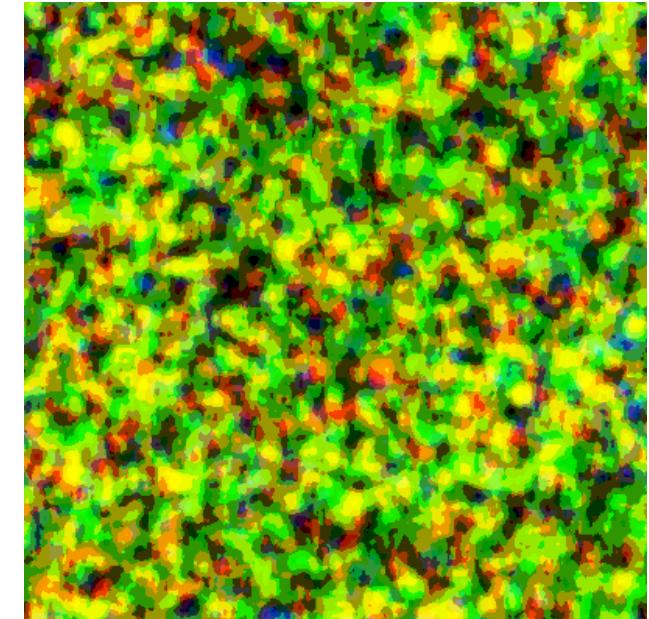
Suppose we want to smooth an image

Replace each pixel by the mean of the surrounding pixels

Represent by matrix giving weights
for each pixel in neighbourhood



$$\begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} /25$$



This is an example of a convolution filter
Create new filters by changing the values in the matrix

AGATAGGAAGAGCGGTTCA GCACGGAAATGCCGAGA

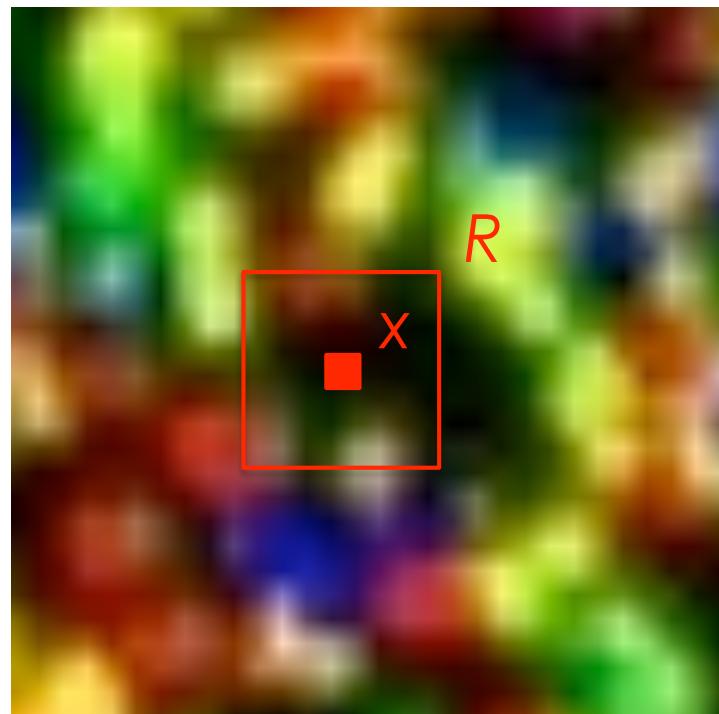
Convolution filters

Take a region around pixel

Multiply every pixel in region by corresponding value in filter F

Sum

$$x_{new} = \mathbf{1}^T (R \circ F) \mathbf{1}$$



E.g. $F = \begin{pmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{pmatrix}, R = \begin{pmatrix} 0.33 & 1.73 & 2.56 \\ 1.18 & 4.70 & 7.36 \\ 2.17 & 6.76 & 10.1 \end{pmatrix}$

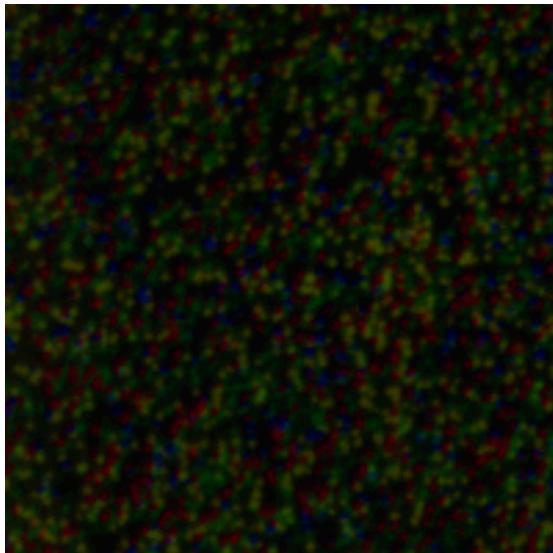
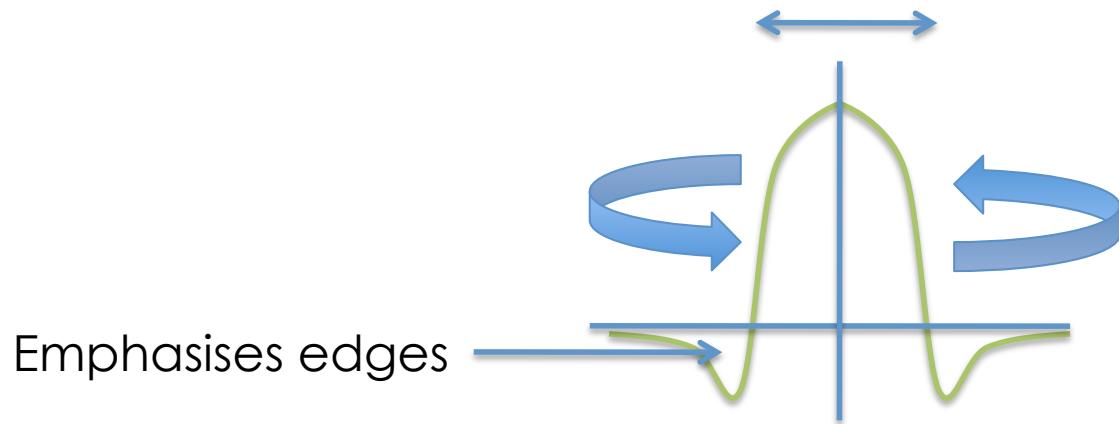
$$\begin{aligned} x_{new} &= \text{sum} \begin{pmatrix} -1 \times 0.33 & -1 \times 1.73 & -1 \times 2.56 \\ -1 \times 1.18 & 8 \times 4.70 & -1 \times 7.36 \\ -1 \times 2.17 & -1 \times 6.76 & -1 \times 10.1 \end{pmatrix} \\ &= 5.41 \end{aligned}$$

Normally do calculations in Fourier space - more efficient

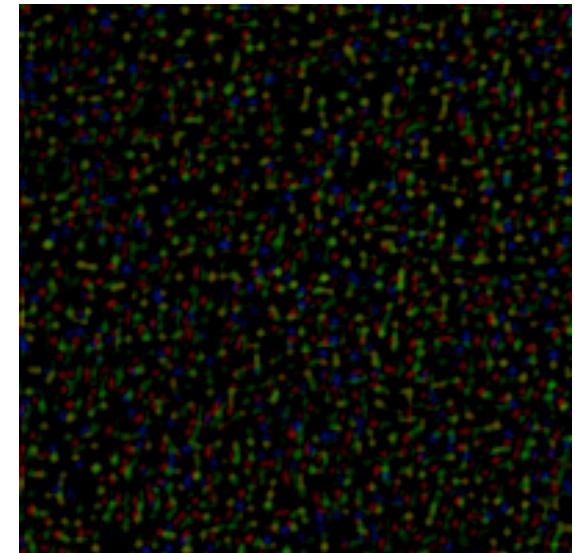
AGATAGGAAGAGCGGTTCA
Filtering

Mexican hat = smoothing + edge detection

Smoothes around central pixel



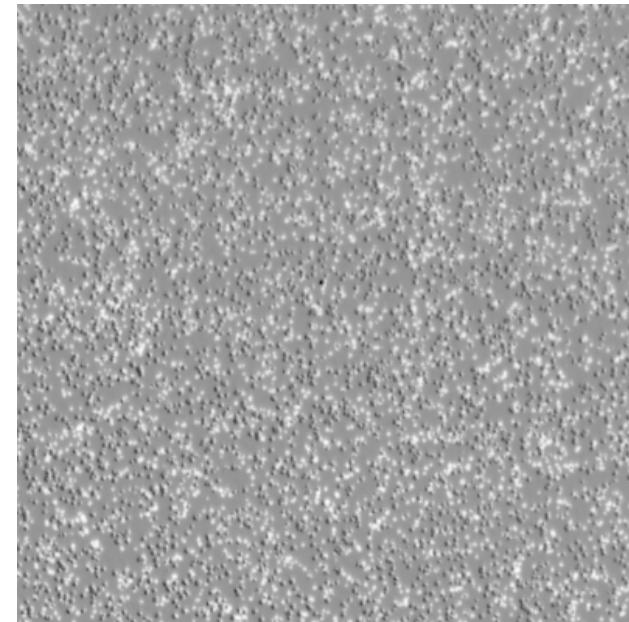
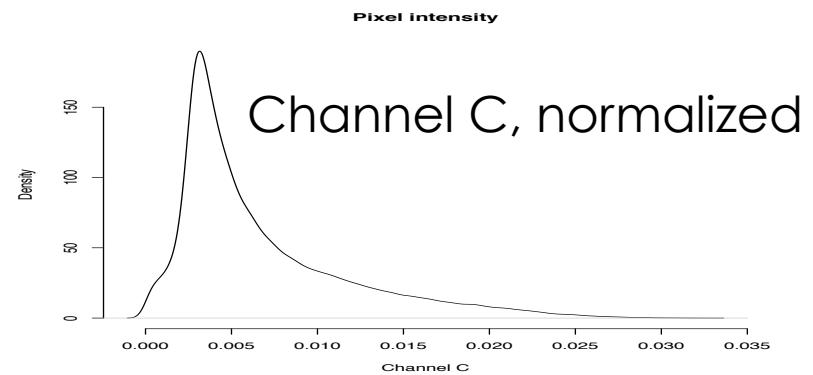
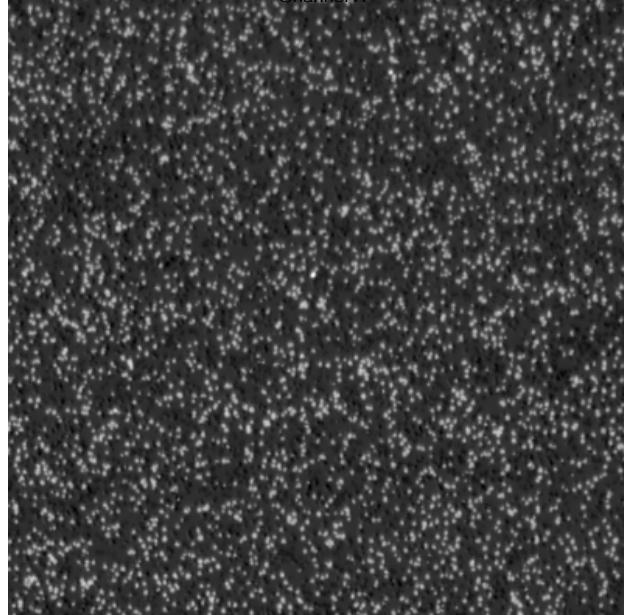
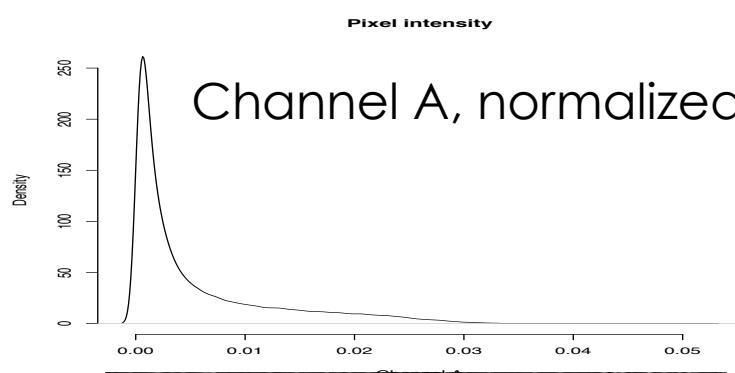
Apply filter
to each channel



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Normalization - background and noise

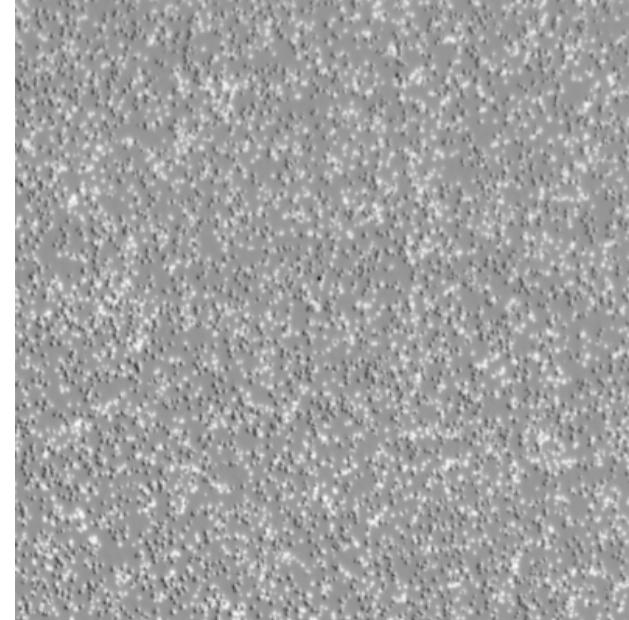
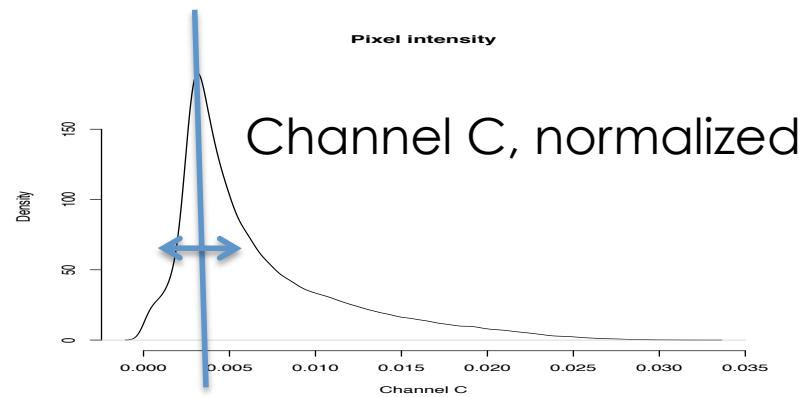
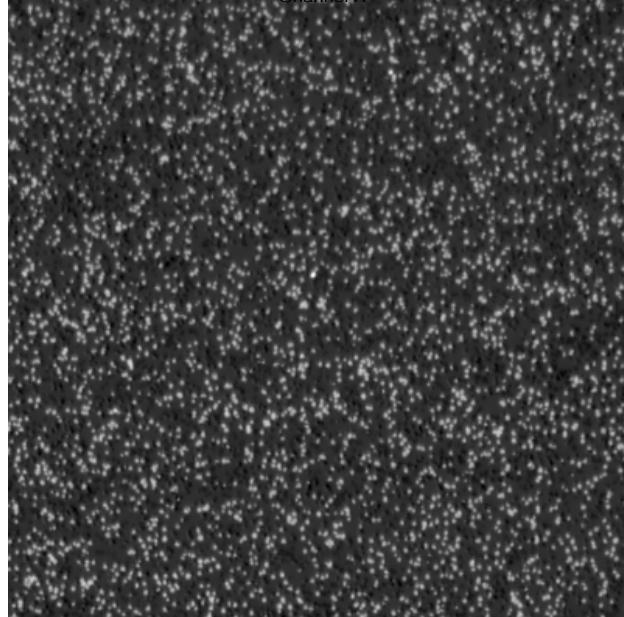
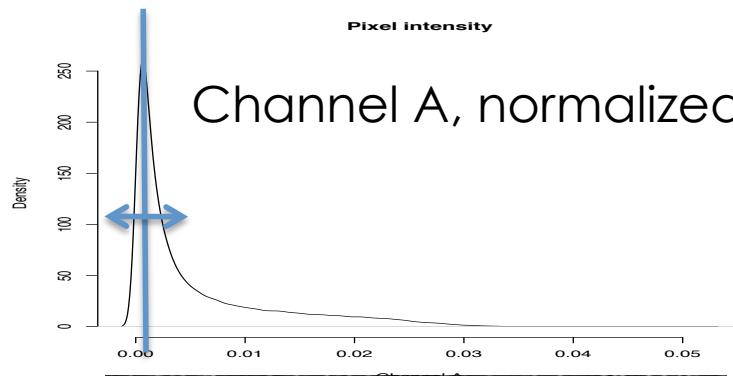
Background fluorescence: flow cell, unincorporated dyes, etc



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Normalization - background and noise

Robust estimates of mean and standard deviation



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Cluster identification

Warning: based on old version of pipeline;
this bit has probably changed more than any other

“Blank slide” model

Background fluorescence

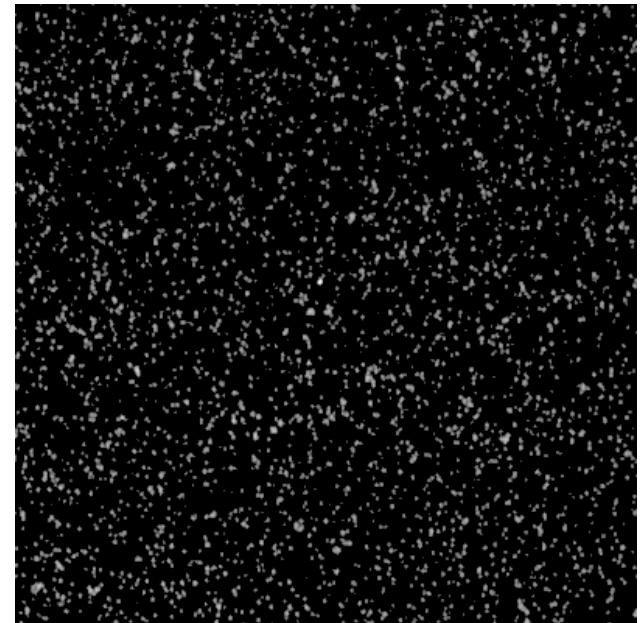
mean

Noise estimate

variance

Keep pixels 4 standard deviations above mean
4 sd ~ Q45 (30 errors per 1 million pixels)

Thresholded tile

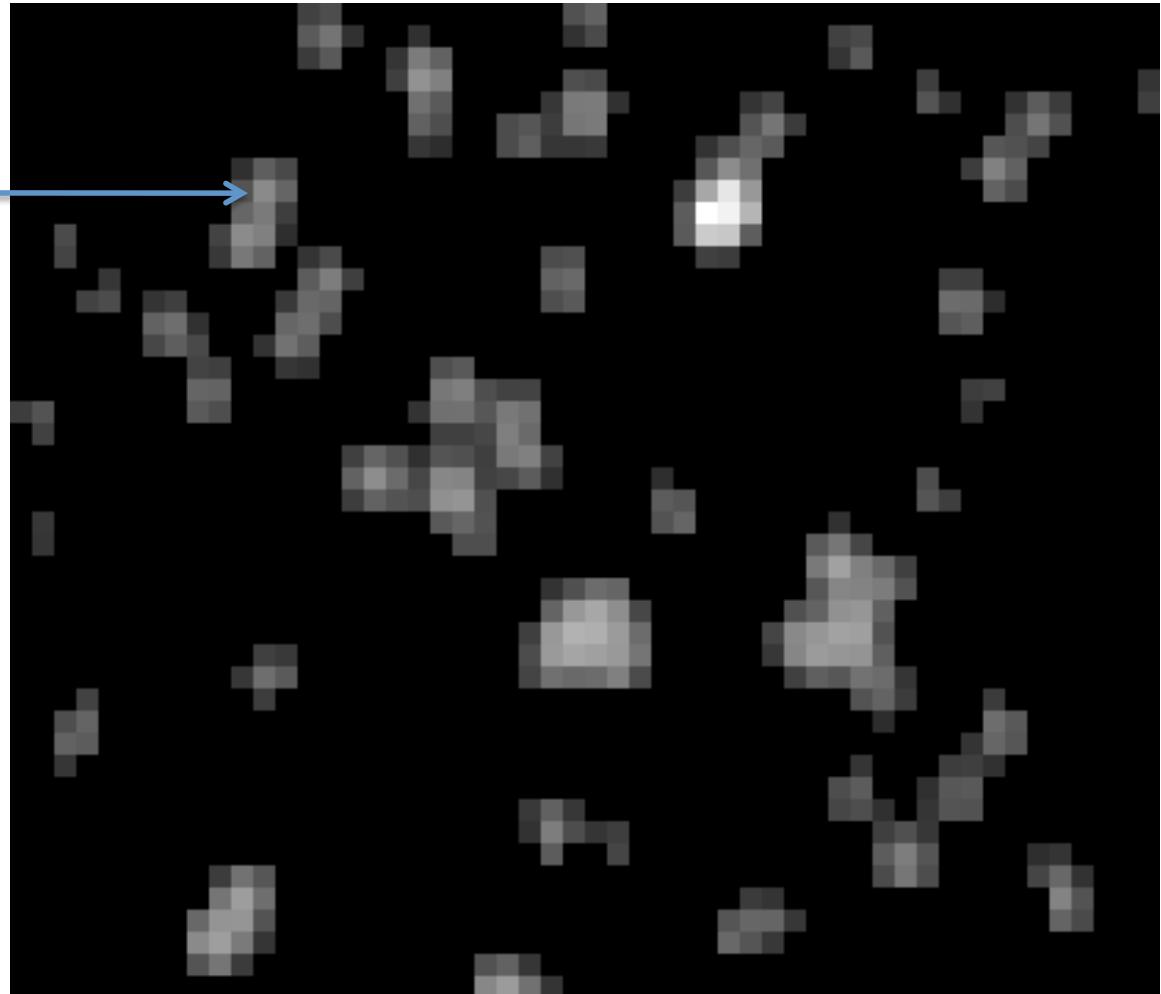


Background and noise estimated in regions

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Cluster identification

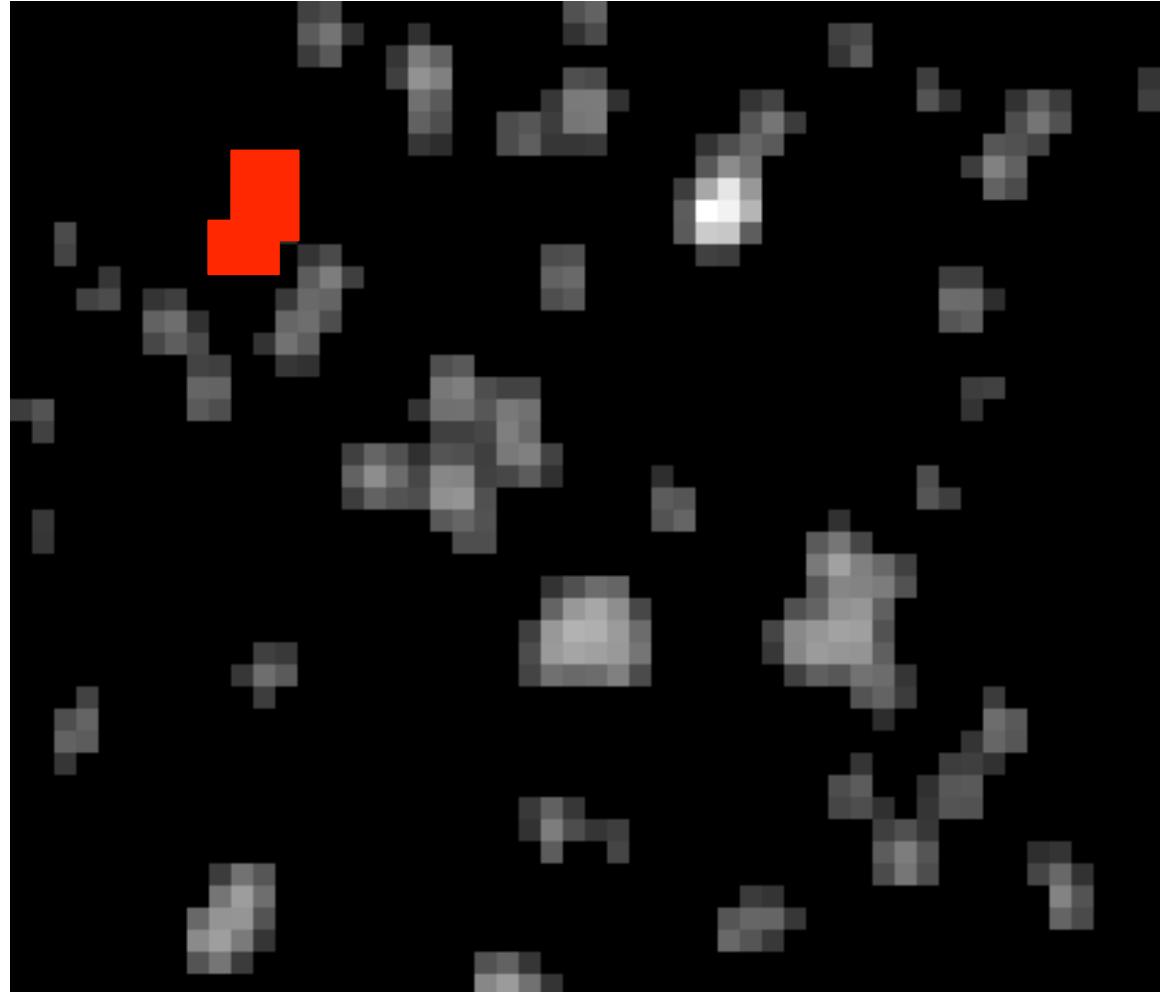
- Find cluster



AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

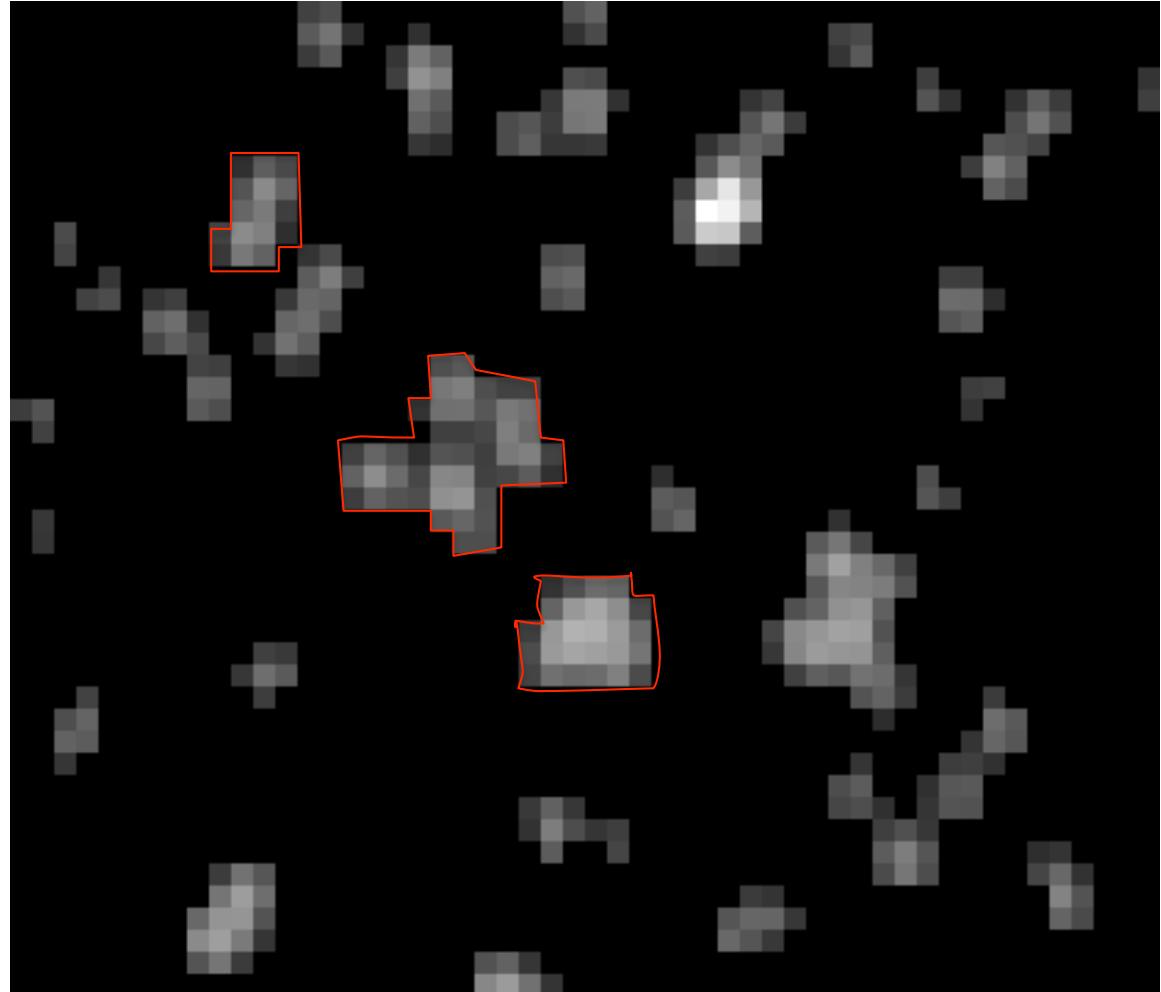
Cluster identification

- Find cluster
- Expand



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA
Cluster identification

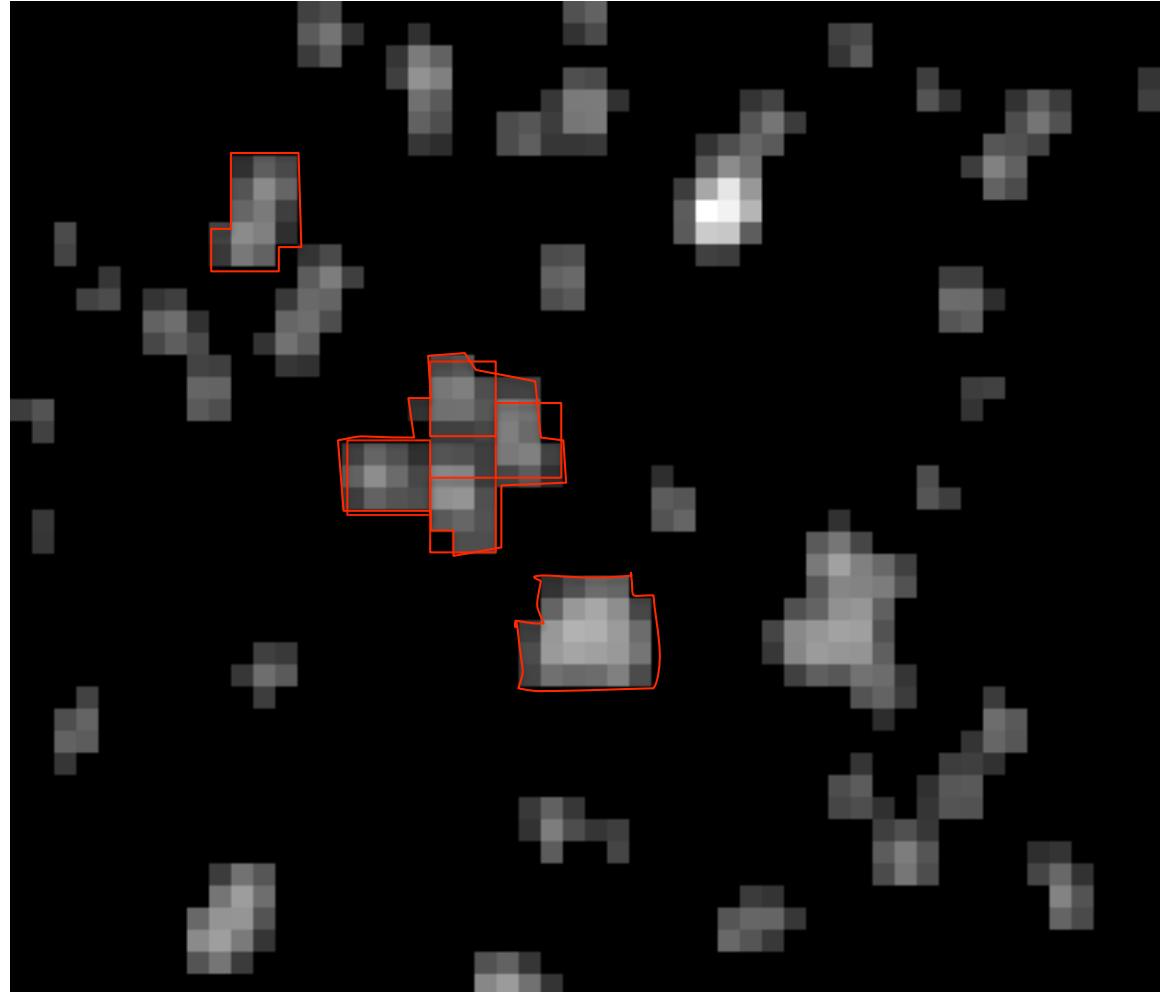
- Find cluster
- Expand
- Find border



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Cluster identification

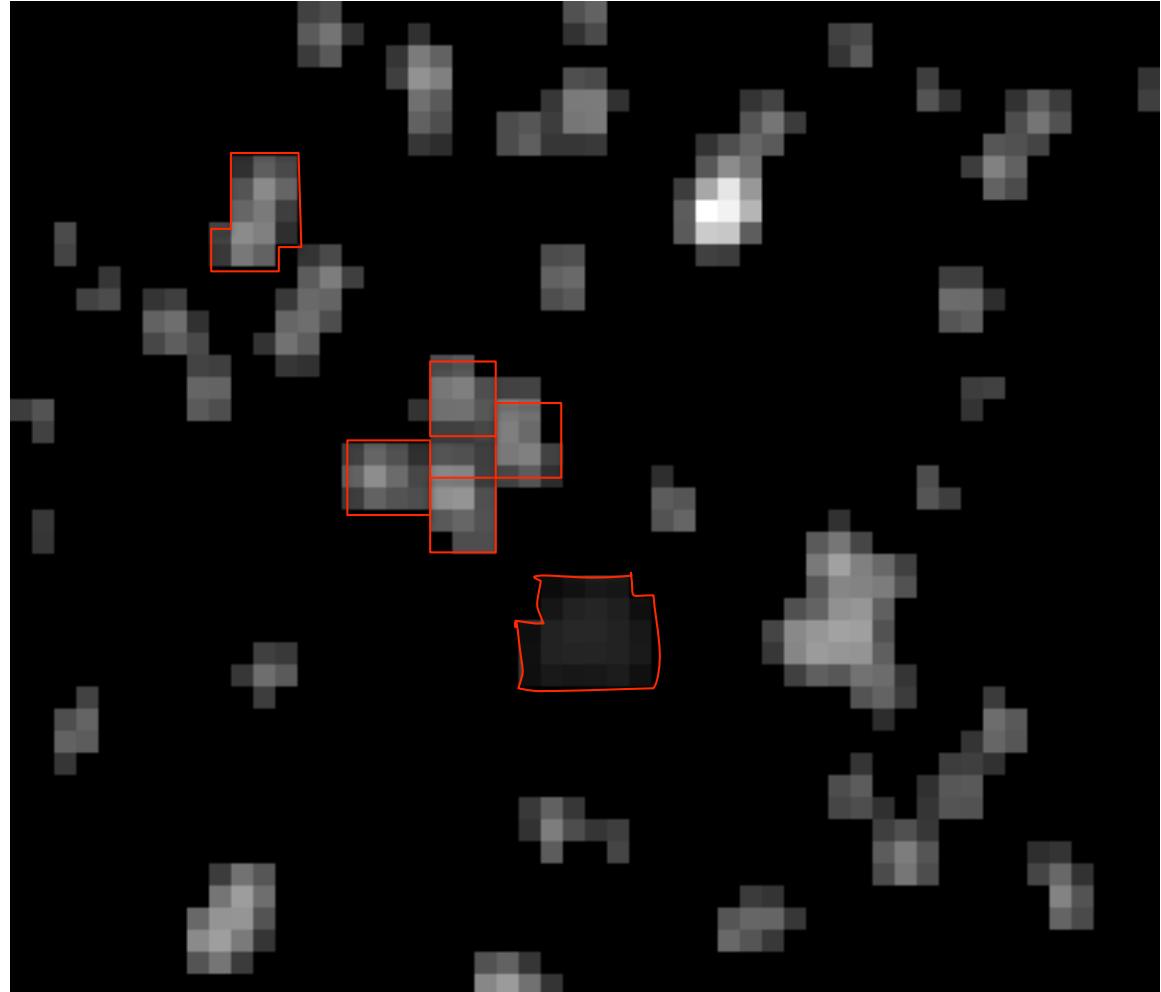
- Find cluster
- Expand
- Find border
- Deblend (split) large clusters



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Cluster identification

- Find cluster
- Expand
- Find border
- Deblend (split) large clusters
- Discard extremely large (probably contamination)



AGATAGGAAGAGCGGTTCA
GCACGGAAATGCCGAGA

Local background

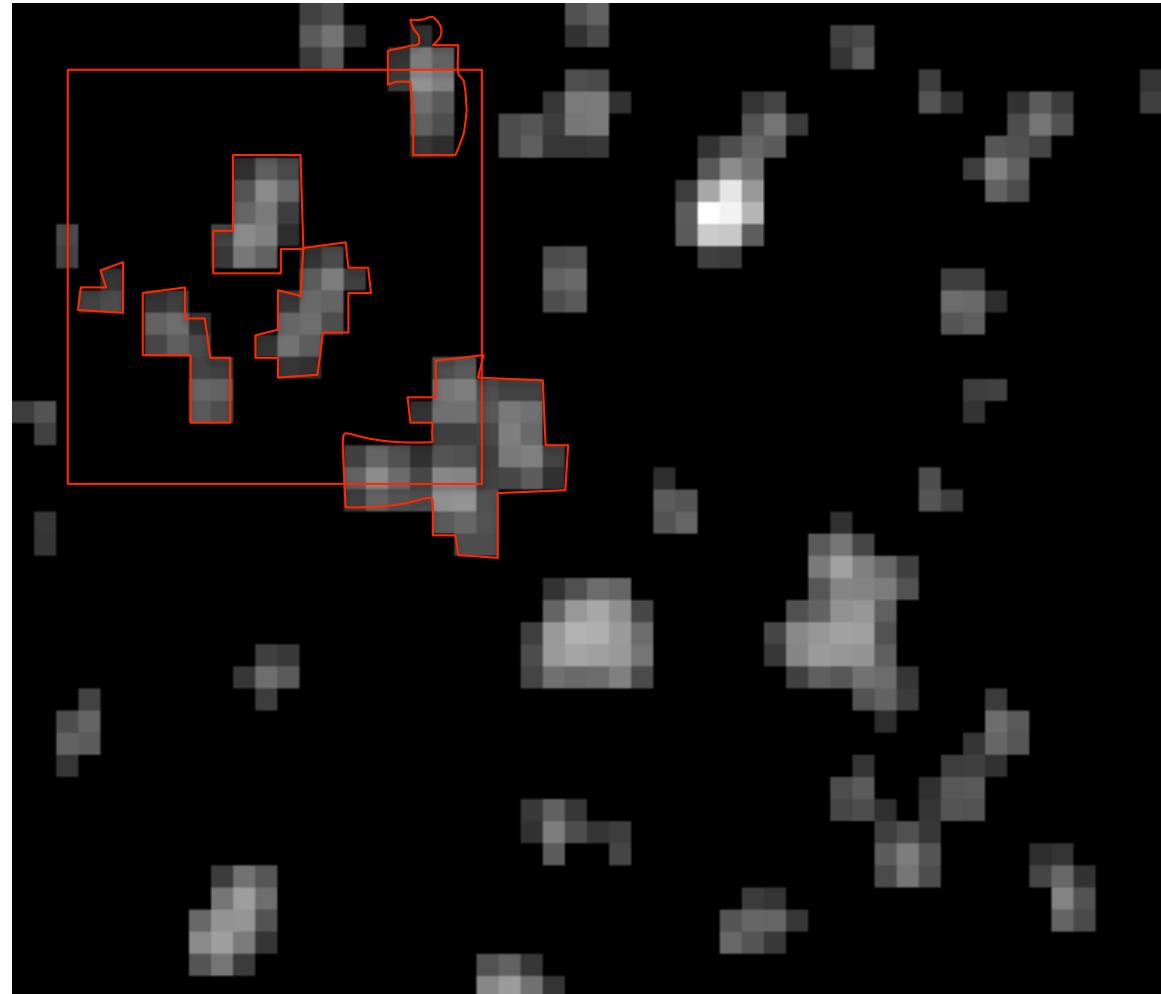
10x10 window
around cluster

Take pixels not part of
any cluster

Calculate new
background noise

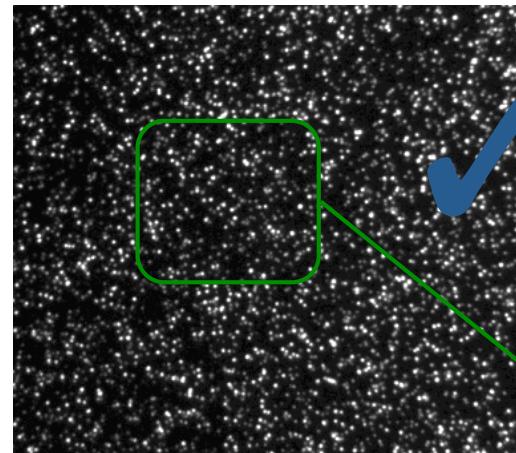
Correct cluster

Find brightest pixel for
base caller



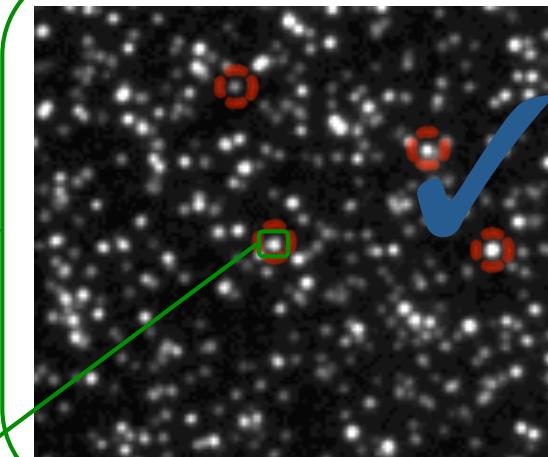
AGATAGGAAGAGCGGTTCA
Analysis pipeline

Raw images

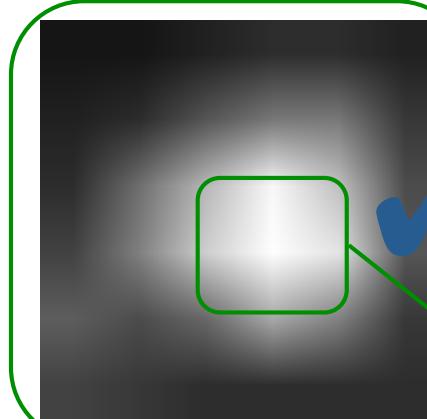


One image per channel, so
four images per cycle

Find clusters



Read intensities

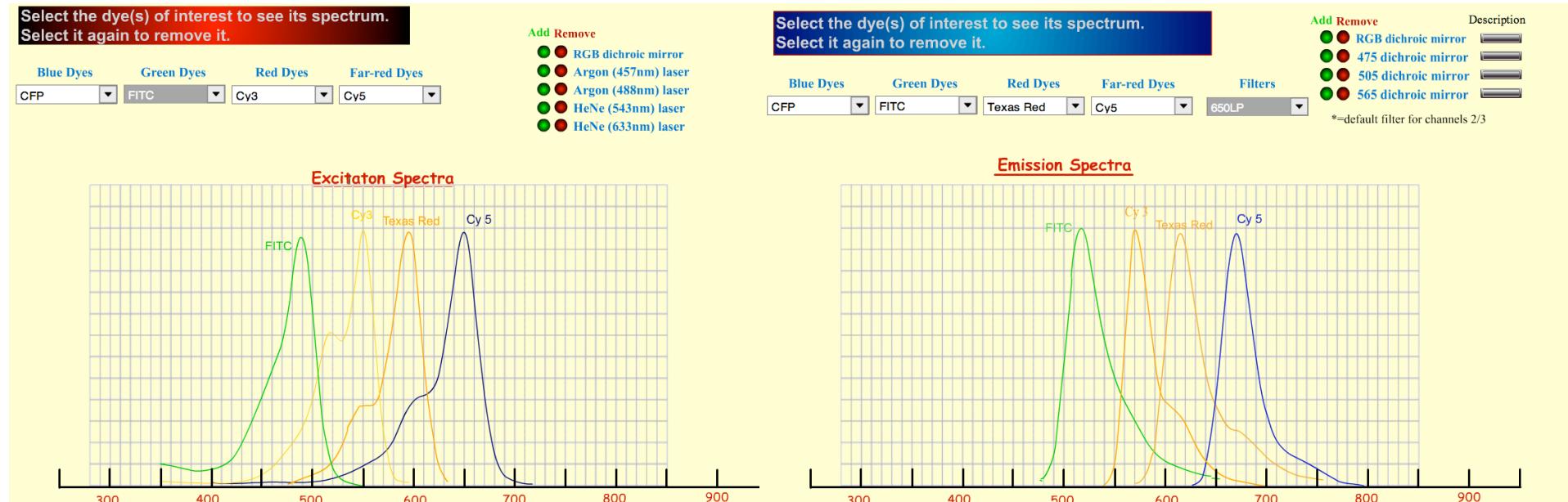


Estimate cross-talk and
phasing, then call bases

#CH4:OBJ13808				
3210.3	3617.2	242.0	519.3	
62.1	128.3	-101.5	2556.5	
125.8	145.9	-34.9	2685.2	
1968.5	1053.9	4.5	59.6	
1732.1	995.9	5.0	48.1	

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Cross Talk

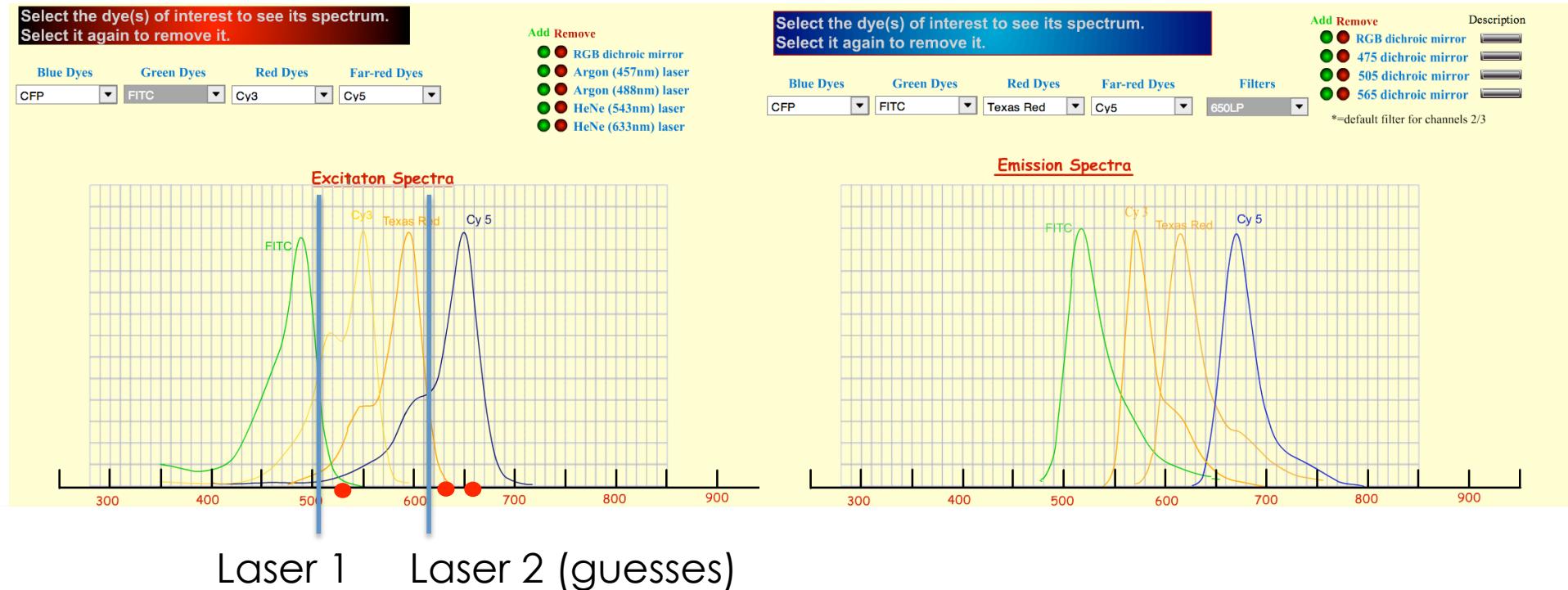


Excitation spectra shows efficiency of wave-length absorption
 Emission spectra shows wave-length of emitted light
 Wave-length of emission ~ independent of absorption

Dyes taken from SOLiD marketing material, with FAM replaced by FITC (excitation and emission spectra not available).
 Spectra from http://www.mcb.arizona.edu/IPC/spectra_page.htm

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Cross Talk



Pick lasers to excite as few fluorophores as possible

- Each putative laser excites two fluorophores
- Laser 1 excites Texas Red and Cy5 a small amount

Illumina uses a three laser system with wave length 532nm, 635nm and 660nm, two of which are used for imaging and one for focus. Shown by red dots

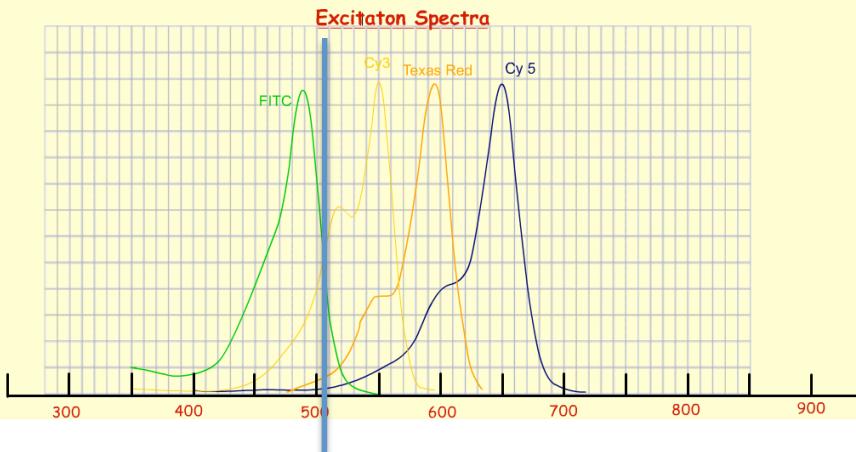
AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Cross Talk

Select the dye(s) of interest to see its spectrum.
Select it again to remove it.

Blue Dyes Green Dyes Red Dyes Far-red Dyes
 CFP FITC Cy3 Cy5

- Add Remove
- RGB dichroic mirror
 - Argon (457nm) laser
 - Argon (488nm) laser
 - HeNe (543nm) laser
 - HeNe (633nm) laser

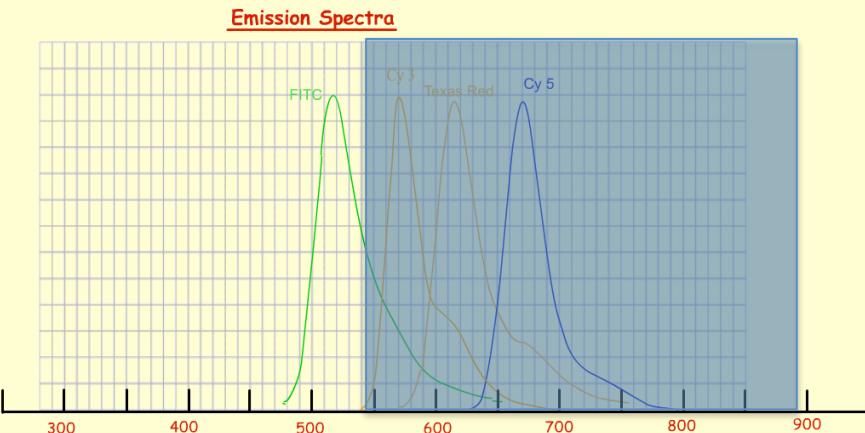


Exciting both FITC and Cy3 with laser -- mixed emission

Select the dye(s) of interest to see its spectrum.
Select it again to remove it.

Blue Dyes Green Dyes Red Dyes Far-red Dyes Filters
 CFP FITC Texas Red Cy5 650LP

- Add Remove
- RGB dichroic mirror
 - 475 dichroic mirror
 - 505 dichroic mirror
 - 565 dichroic mirror
- Description
- *=default filter for channels 2/3



Use a filter to block Cy3 wave lengths, so observed signal is pure

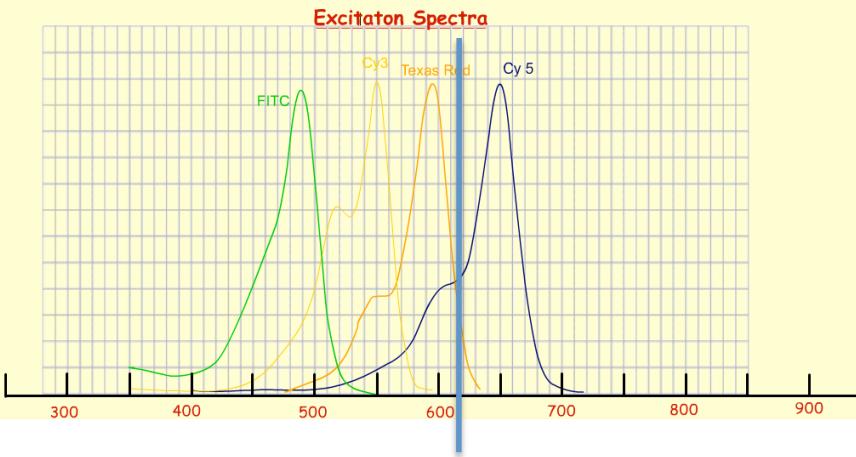
AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Cross Talk

Select the dye(s) of interest to see its spectrum.
Select it again to remove it.

Blue Dyes Green Dyes Red Dyes Far-red Dyes
 CFP FITC Cy3 Cy5

- Add Remove
- RGB dichroic mirror
 - Argon (457nm) laser
 - Argon (488nm) laser
 - HeNe (543nm) laser
 - HeNe (633nm) laser

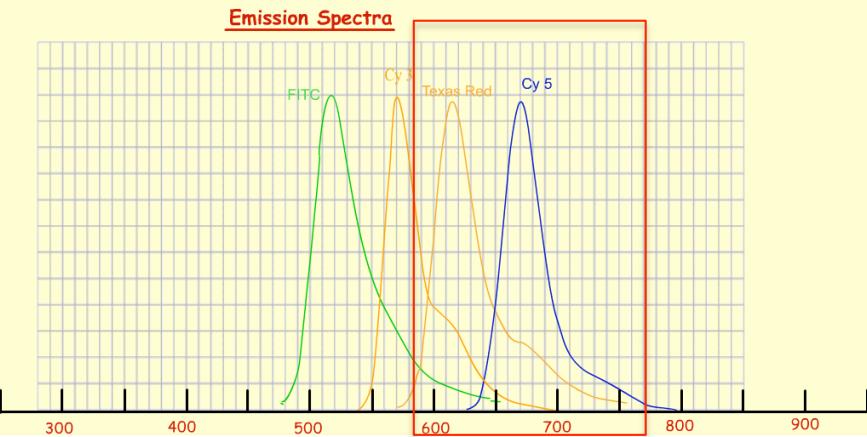


Exciting both Texas Read
and Cy5 with laser -- mixed
emission

Select the dye(s) of interest to see its spectrum.
Select it again to remove it.

Blue Dyes Green Dyes Red Dyes Far-red Dyes Filters
 CFP FITC Texas Red Cy5 650LP

- Add Remove
- RGB dichroic mirror
 - 475 dichroic mirror
 - 505 dichroic mirror
 - 565 dichroic mirror
- Description
- *=default filter for channels 2/3



Emission spectra have strong overlap,
hard to construct filter to only allow
one through

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

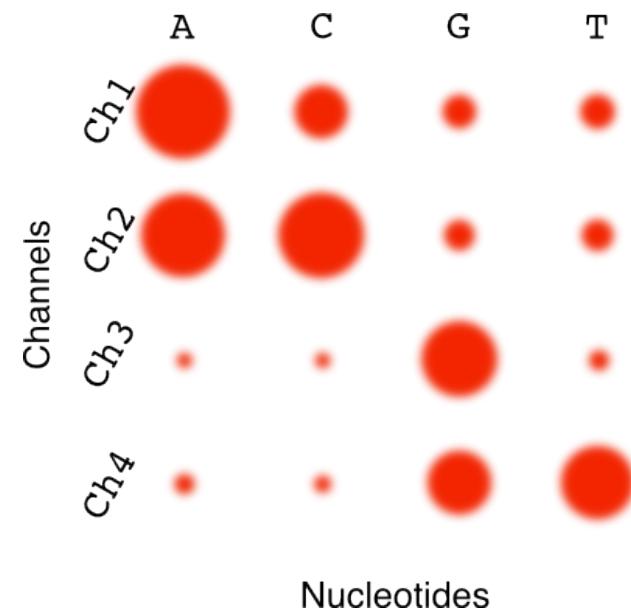
Cross talk

Channel = specific combination of laser and filter

Observe channels rather than nucleotides

Represent cross talk by a matrix

Entries represent how bright
each fluorophore appears in
each channel



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

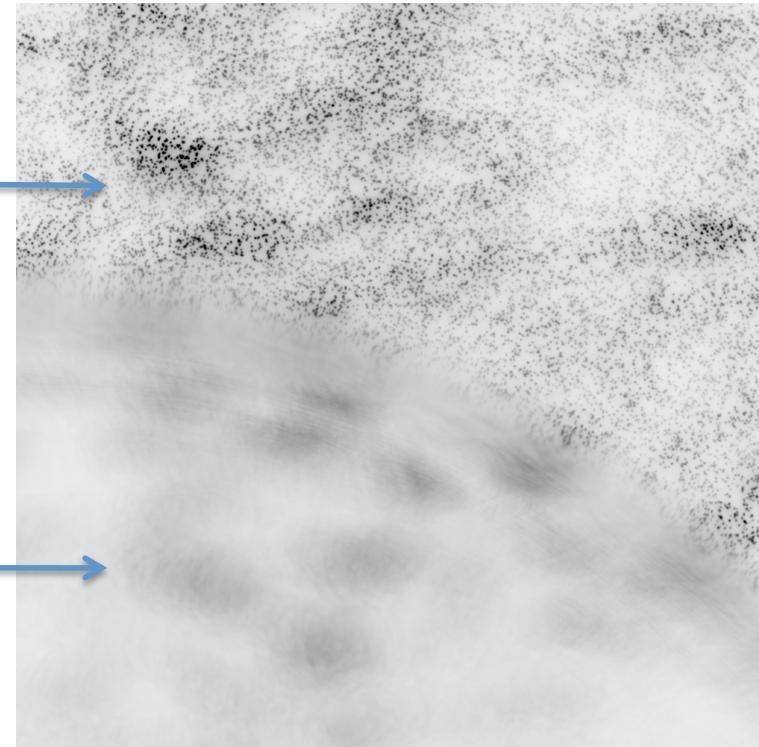
Cross Talk

Laser = coherent light,

Regular patterns of light and dark depending on wavelength

Use a mode scrambler to even out

Mode scrambler problems,
bright and dark patches

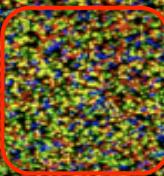
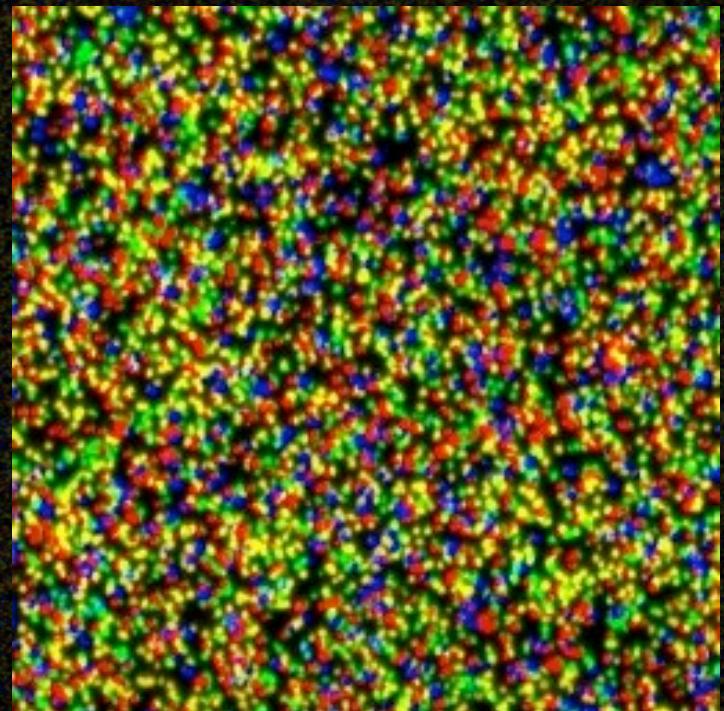


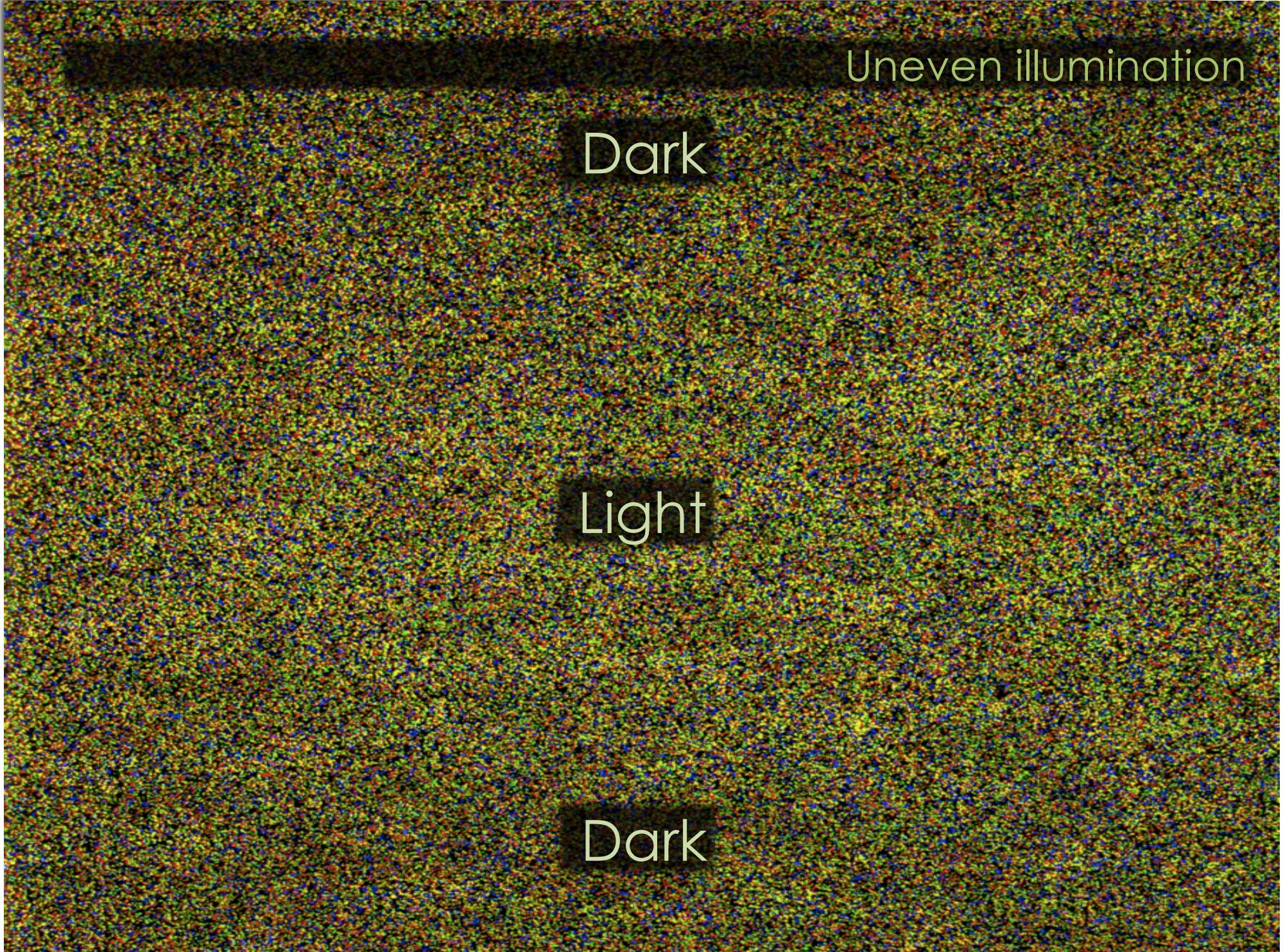
Bubble in flow cell,
All clusters lost here for
this cycle



Image courtesy of James Bonfield, Sanger Institute

False colour image of first cycle, crosstalk corrected





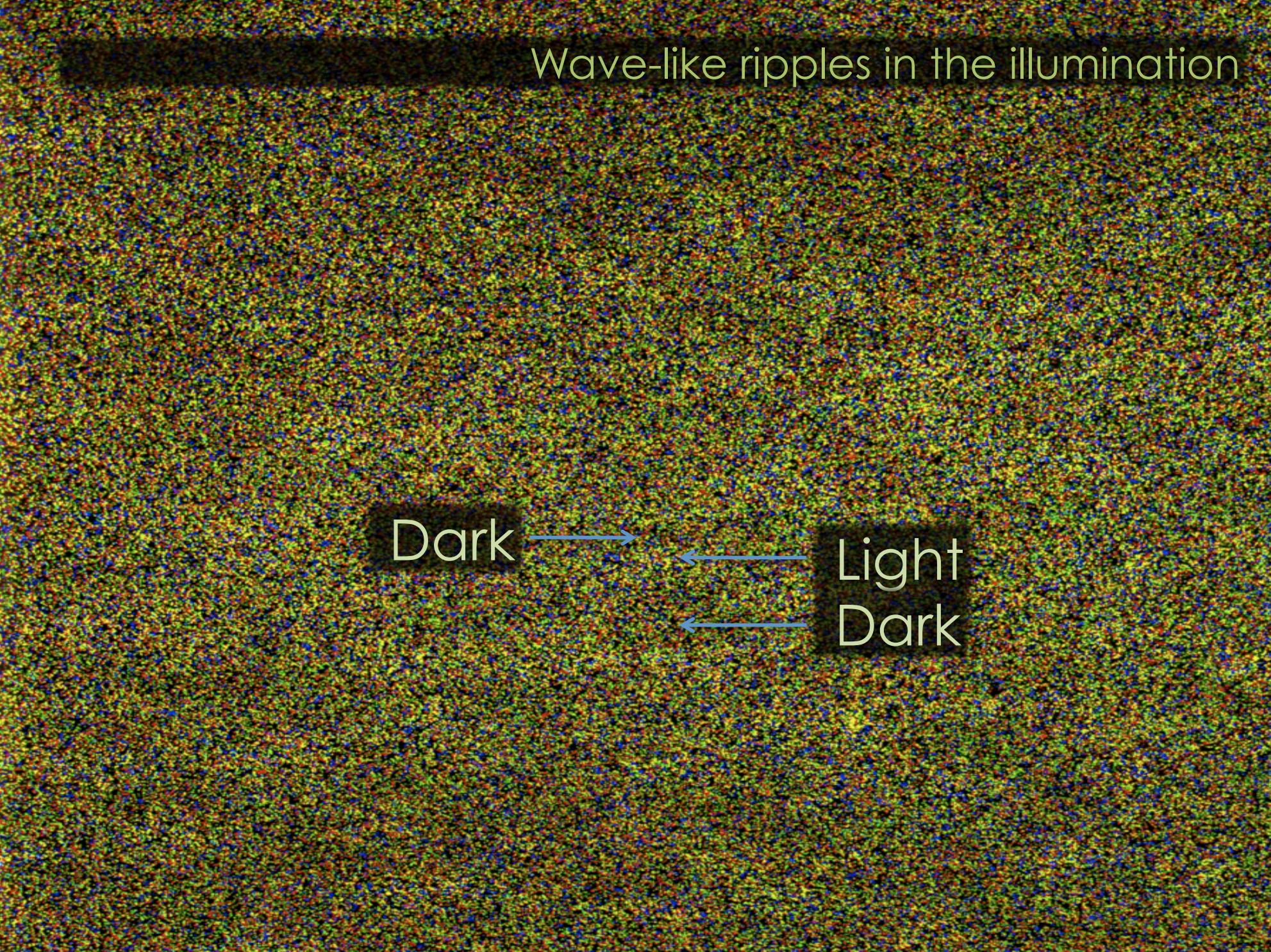
Uneven illumination

Dark

Light

Dark

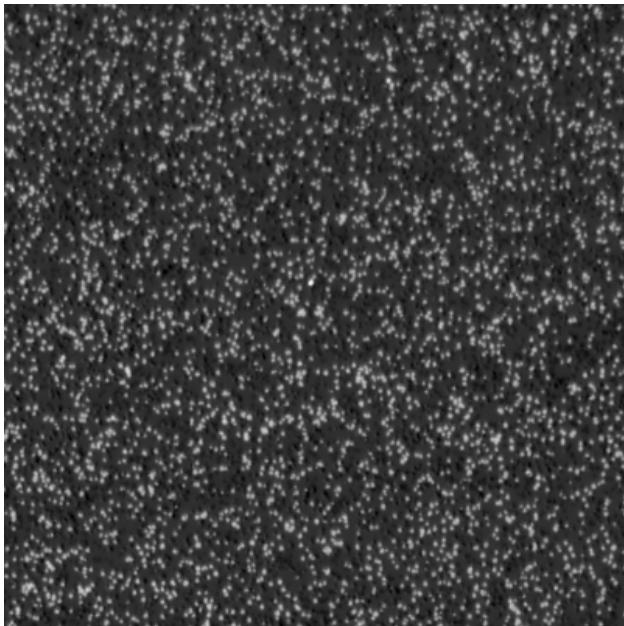
Wave-like ripples in the illumination



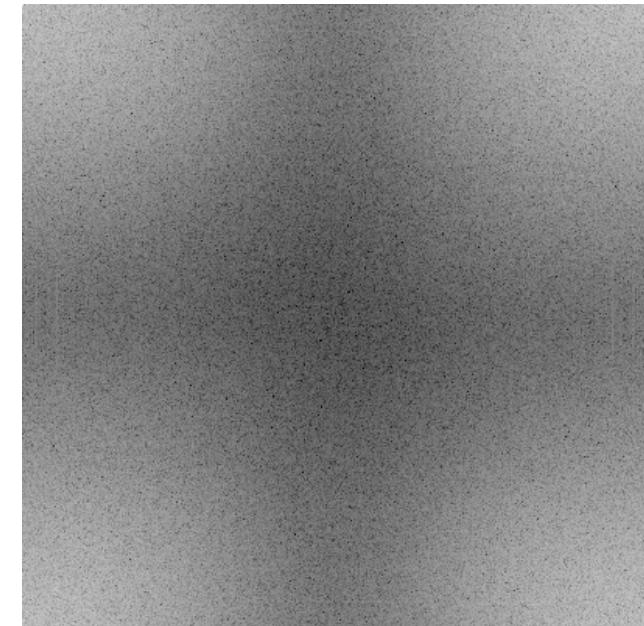
Dark → ← Light
Dark

AGATAGGAAGAGCGGTTCACCACGGAAATGCCGAGA
Variation in luminescence

Original image



log Mod FT



Fourier transform

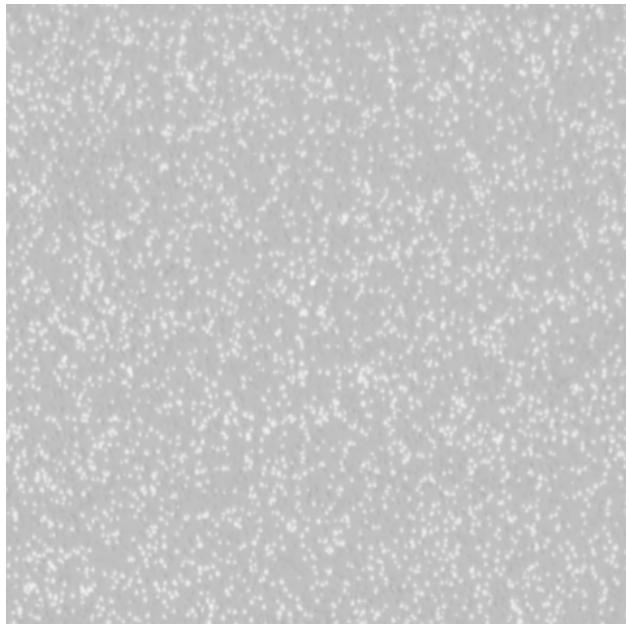


AGATAGGAAGAGCGGTTCACCACGGAAATGCCGAGA

Variation in luminescence

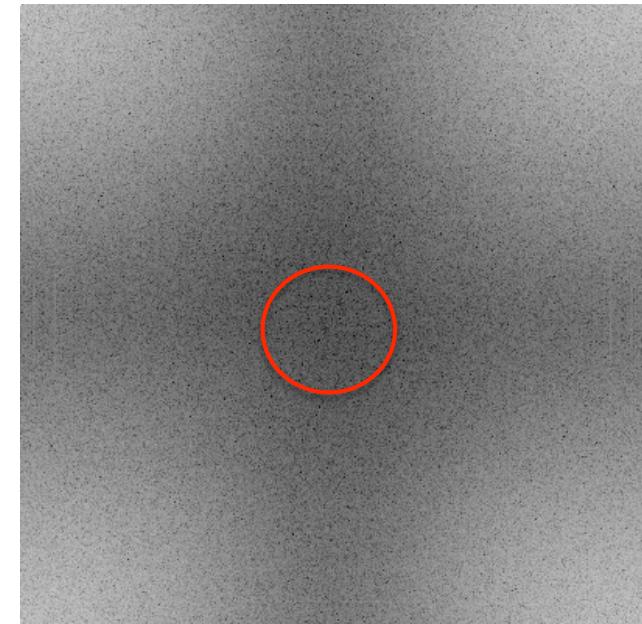
Intensity changes slowly compared to presence / absence of cluster

Original image



Low pass filter
Keep only slowly
varying changes

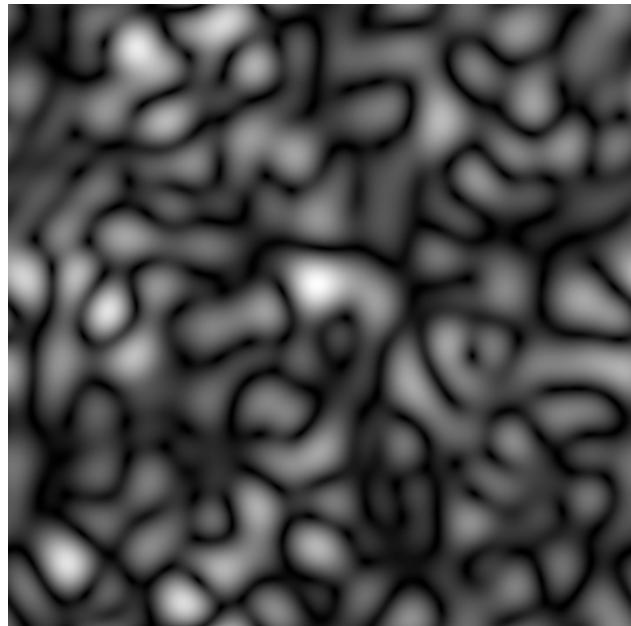
log Mod FT



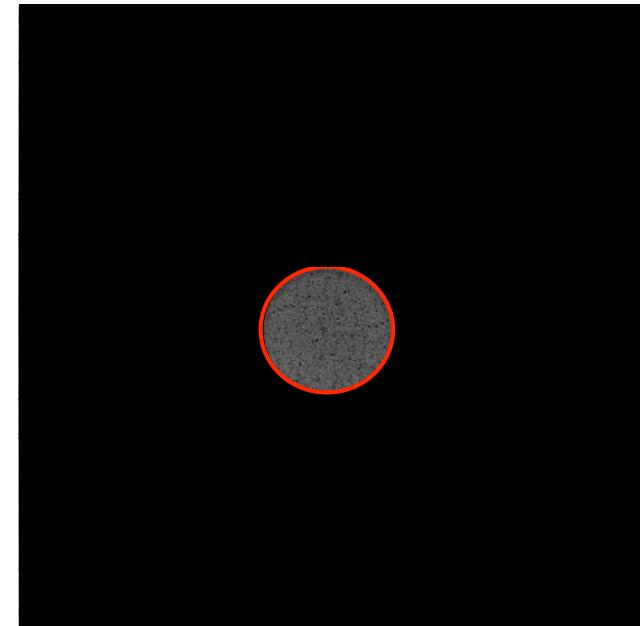
“optimal filtering”
- a step function

AGATAGGAAGAGCGGTTCACCACGGAAATGCCGAGA
Variation in luminescence

Filtered image



log Mod FT



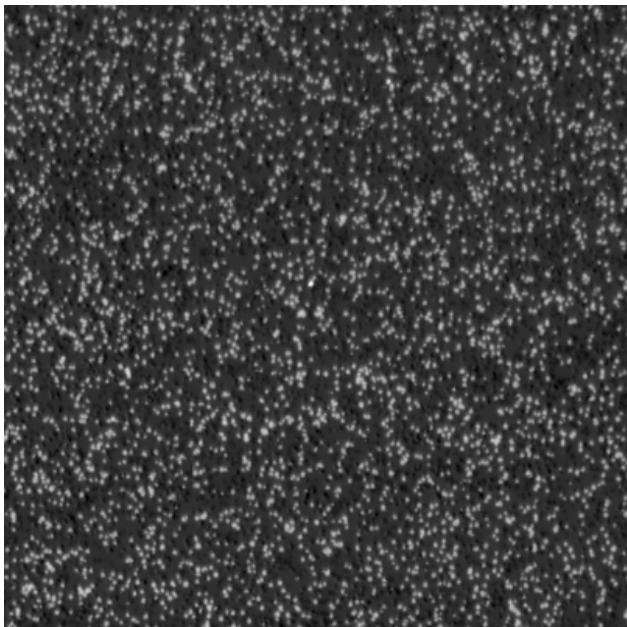
Fourier transform



AGATAGGAAGAGCGGTTCACCACGGAAATGCCGAGA
Variation in luminescence

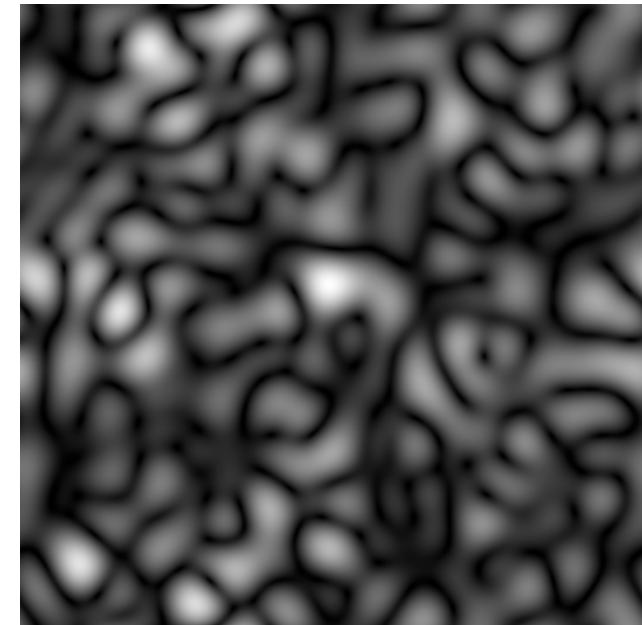
Channel A

IQR: -3.5×10^{-5} -- 4.8×10^{-3}



Filtered, normalized

IQR: 3.1×10^{-6} -- 8.2×10^{-6}



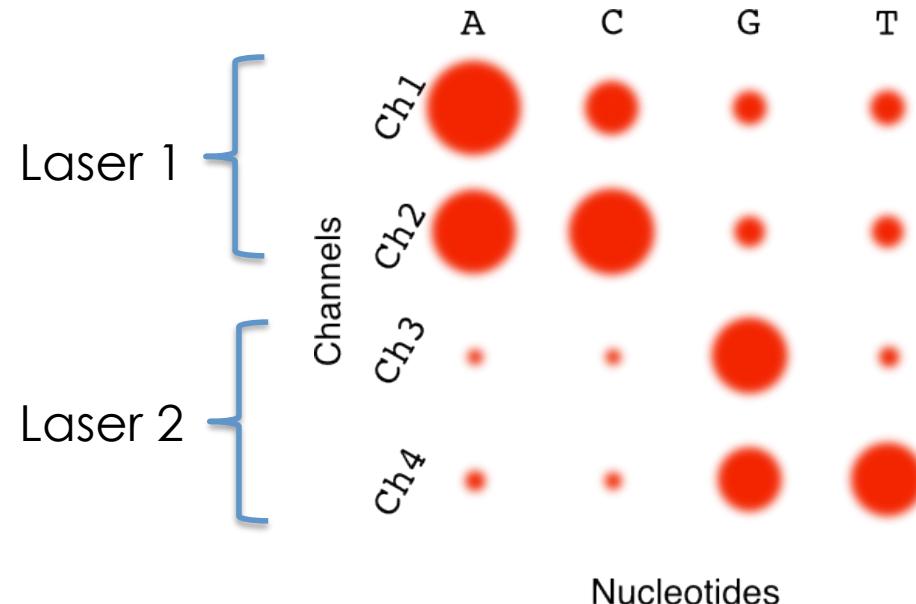
Normalized -
accentuate differences

AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

Cross Talk

Variation in laser intensity across flow cell

- three different lasers, different variation in intensity
- variation in cross talk



Variation between cycles/tiles

- Laser warming up, becomes more efficient
- Changes in focus
- Changes in mode scrambler
- Background fluorescence

Effects mostly ignored

AGATAGGAAGAGCGGTTCA
GCACGGAAATGCCGAGA

Phasing

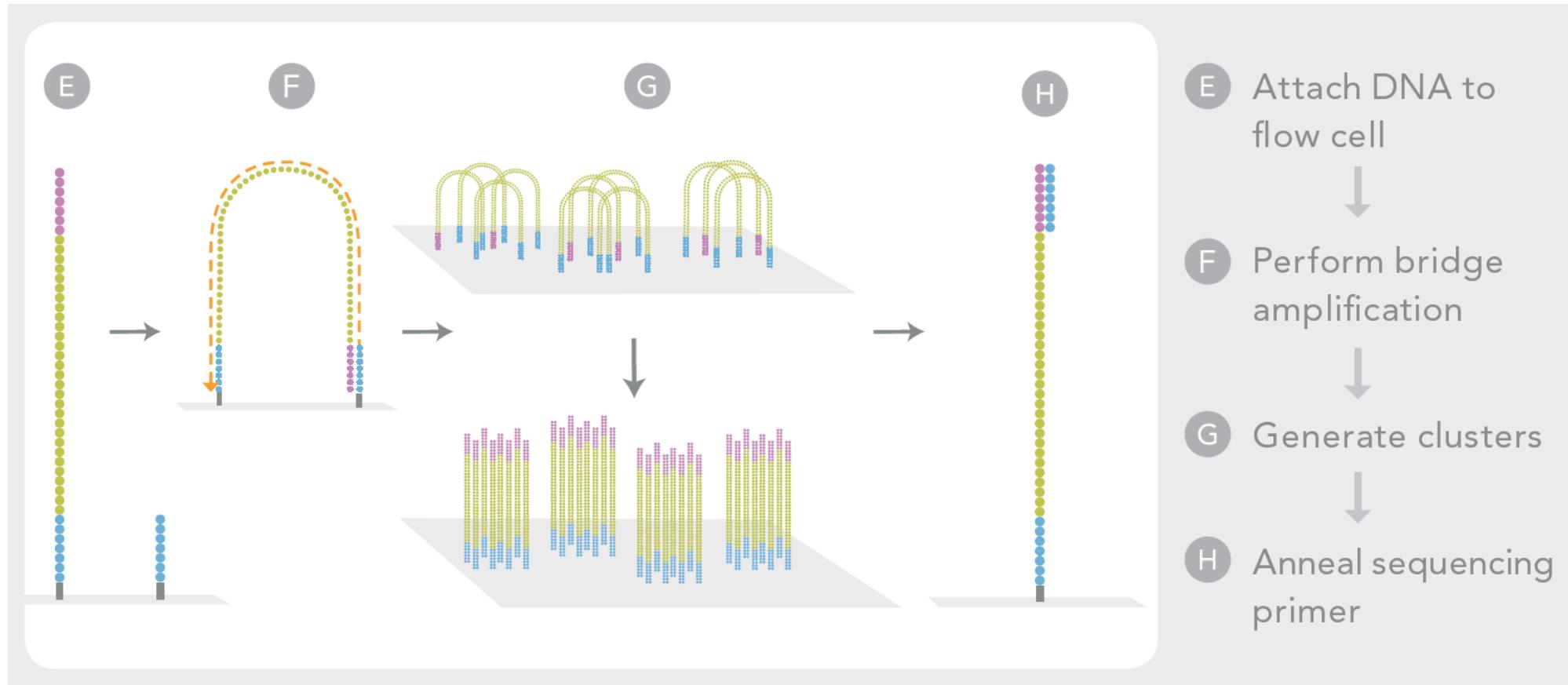
Tendency for molecules to get out of step with others in cluster

Signal from cluster becomes a mixture of previous and future bases

Blurs and becomes harder to tell what current base is

Primarily a chemistry problem

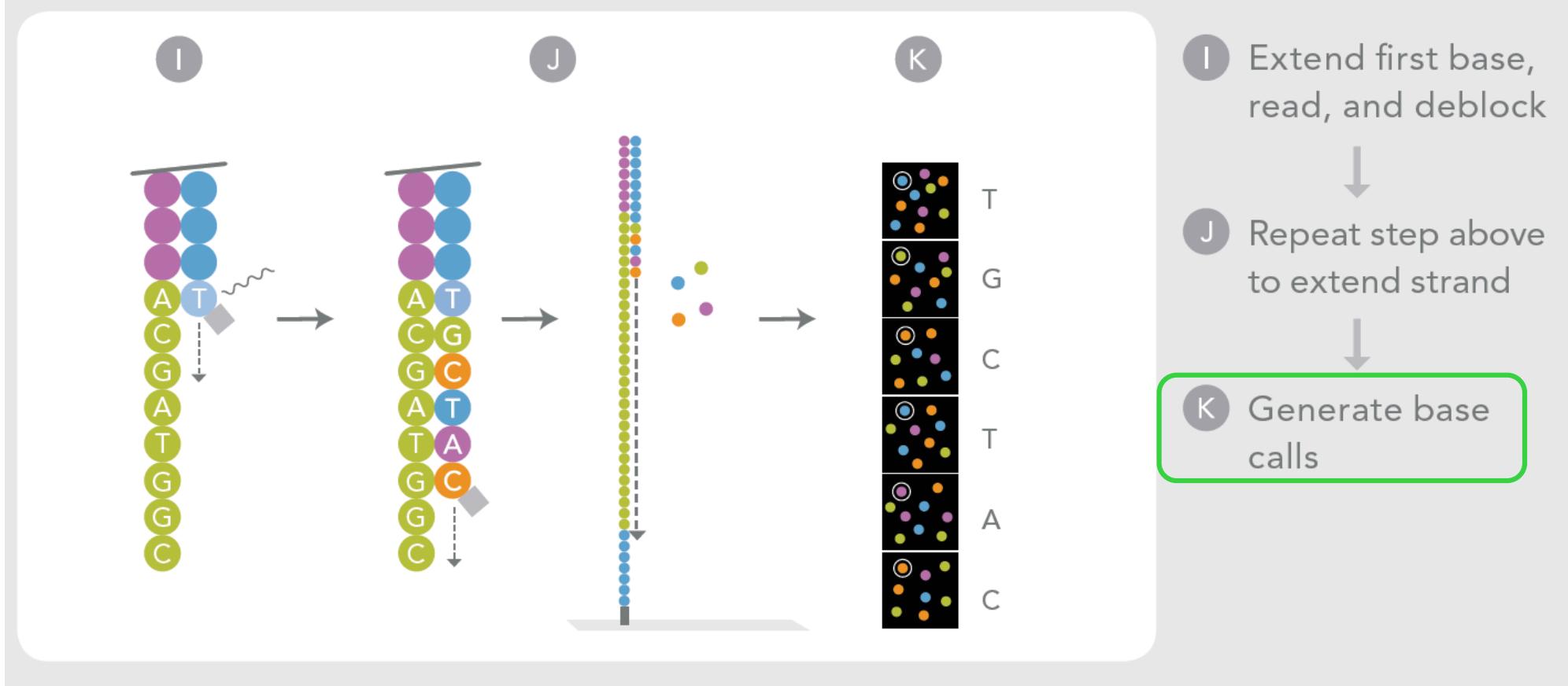
AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA
Illumina chemistry



Source: **illumina** www.illumina.com/sequencing

AGATAGGAAAGAGCGGTTCAGCAGGAATGCCGAGA

Chemistry/Physics

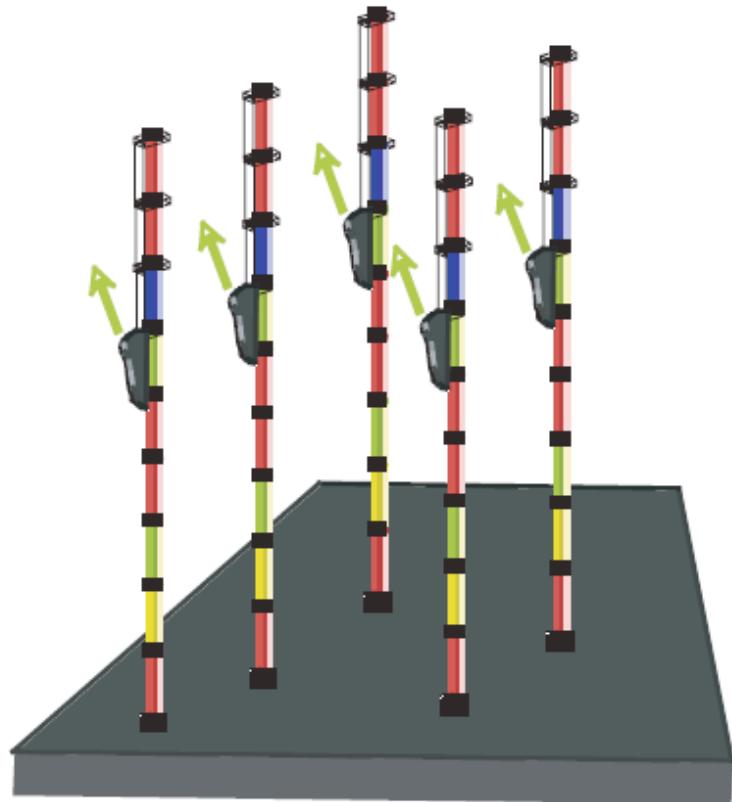


Source:  www.illumina.com/sequencing

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Ideal data

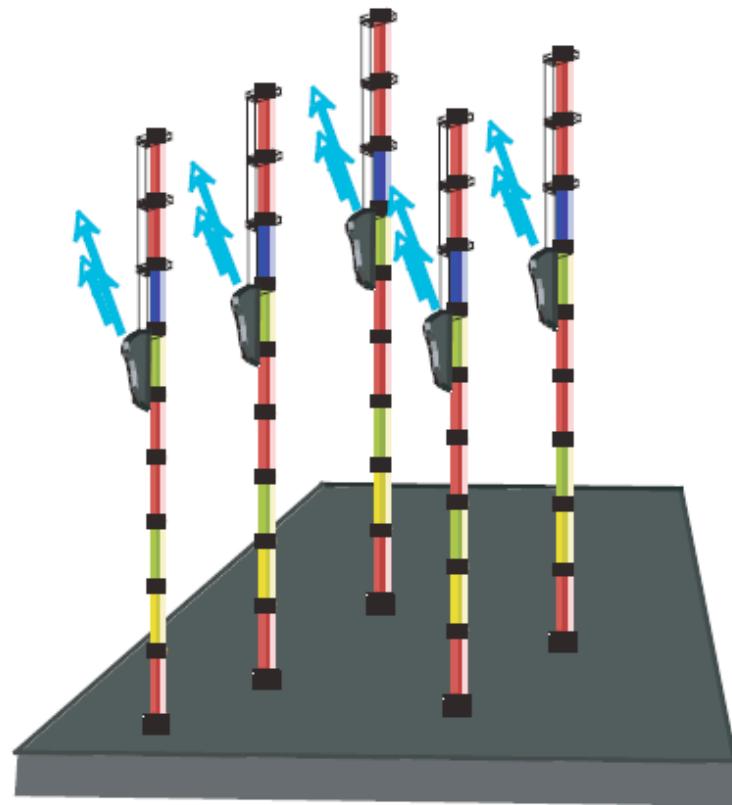
Ideally, signal is strong
(green arrows)



Source:  Erlich et al. (2008) Nature Methods **5**:679–682

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Real data

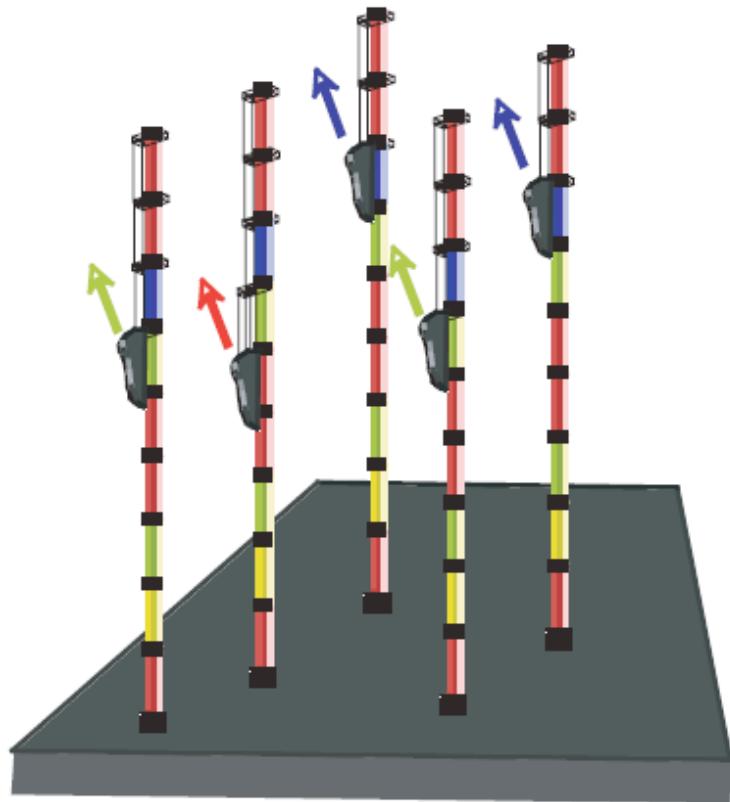


Laser cross-talk:
changes in measured
light emissions, leading
to distorted signal (blue
arrows)

Source:  Erlich et al. (2008) Nature Methods **5**:679–682

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Real data



Phasing:

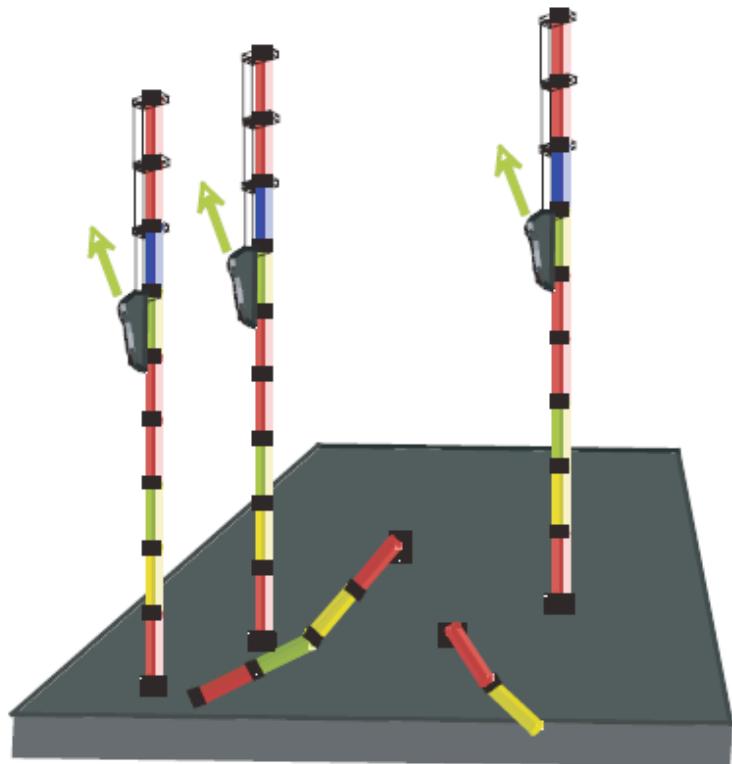
some strands lead (red) or lag behind (blue), leading to mixed signal

Source:  Erlich et al. (2008) Nature Methods **5**:679–682

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA

Real data

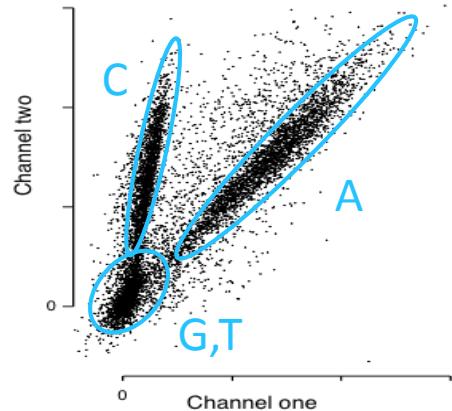
Fading/dimming:
some strands 'die', leading
to reduced signal



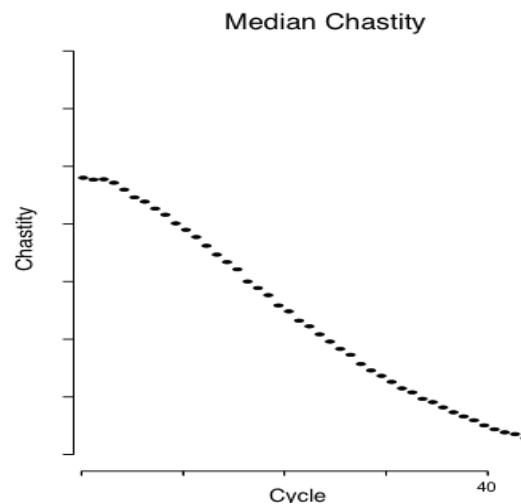
Source:  Erlich *et al.* (2008) *Nature Methods* **5**:679–682

AGATAGGAAGAGCGGTTCAGGAAATGCCGAGA
Sources of error

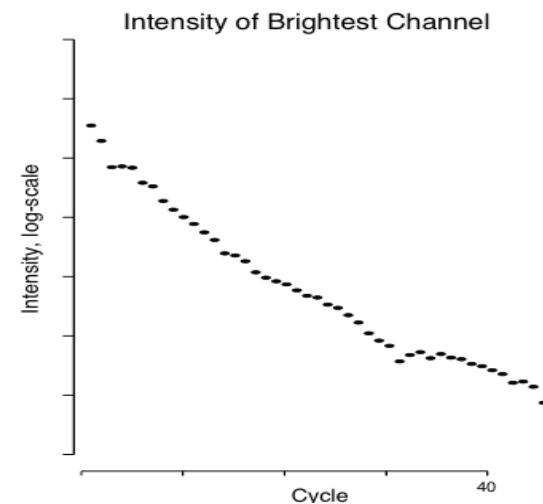
laser cross-talk



phasing



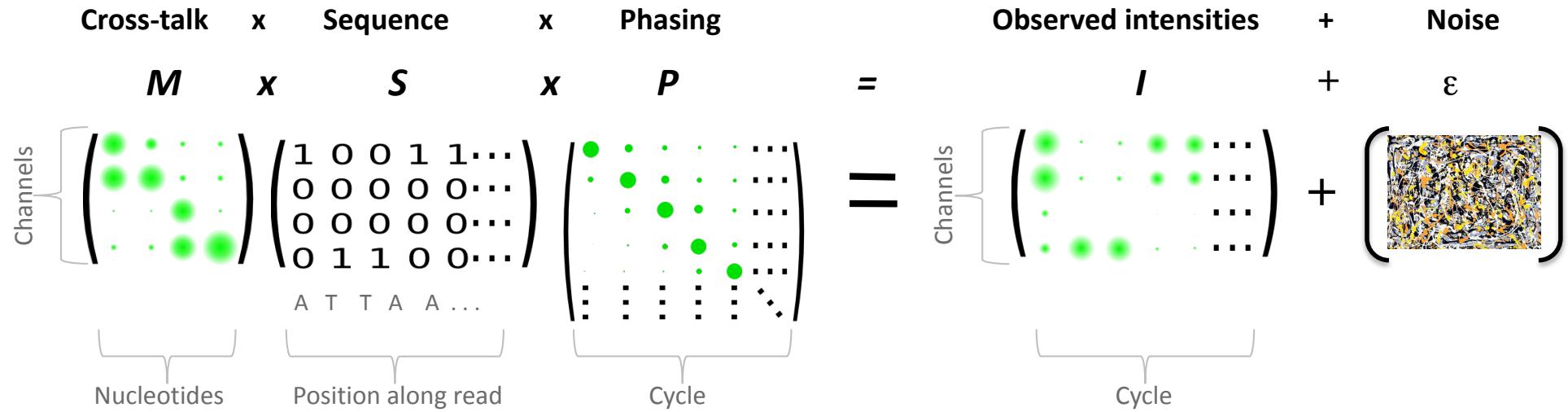
dimming



- + contamination
- + flow cell artefacts
- + random error

AGATAGGAAGAGCGGTCAGCACGGAAATGCCGAGA

AYB statistical model



Hidden linear relationship

$$\text{Vec}(I) = (P^T \otimes M) \text{Vec}(S) + \text{Vec}(\varepsilon)$$

$\text{Vec}(\downarrow\downarrow\downarrow)=$

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Sequence inference

$$\begin{pmatrix} M & x \\ \vdots & \vdots \\ \vdots & \vdots \\ \vdots & \vdots \end{pmatrix}
 \begin{pmatrix} 1 & 0 & 0 & 1 & 1 & \dots \\ 0 & 0 & 0 & 0 & 0 & \dots \\ 0 & 0 & 0 & 0 & 0 & \dots \\ 0 & 1 & 1 & 0 & 0 & \dots \end{pmatrix}
 \begin{pmatrix} x & P \\ \vdots & \vdots \\ \vdots & \vdots \\ \vdots & \vdots \end{pmatrix}
 = \begin{pmatrix} I & \varepsilon \\ \vdots & \vdots \\ \vdots & \vdots \\ \vdots & \vdots \end{pmatrix}
 + \begin{pmatrix} \varepsilon \\ \vdots \\ \vdots \\ \vdots \end{pmatrix}$$

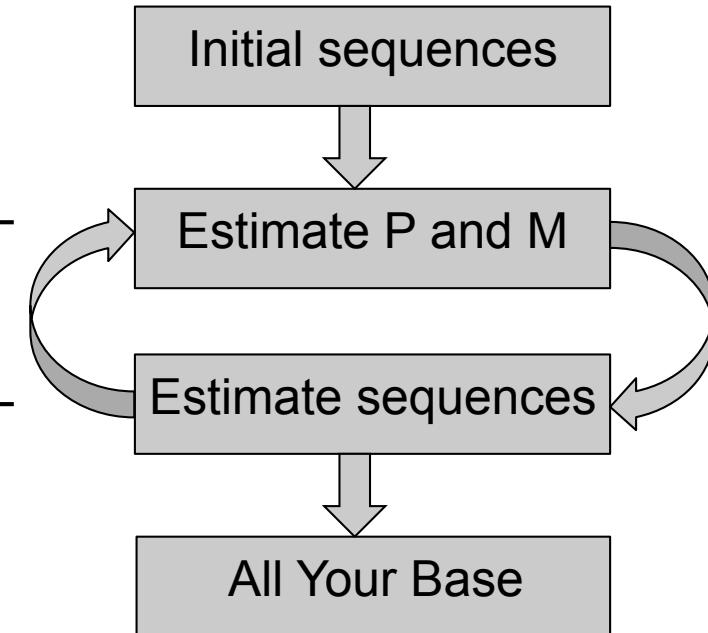
Ordinary Linear Model

noise assumed independent

General Linear Model

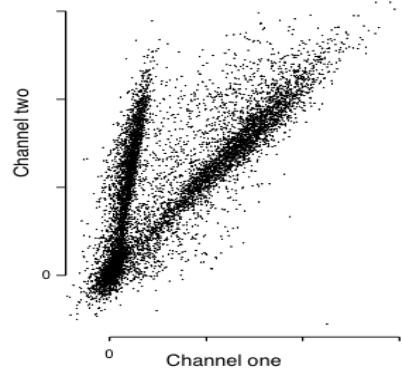
for each cluster

$\text{Var}(noise)$ estimated using all clusters

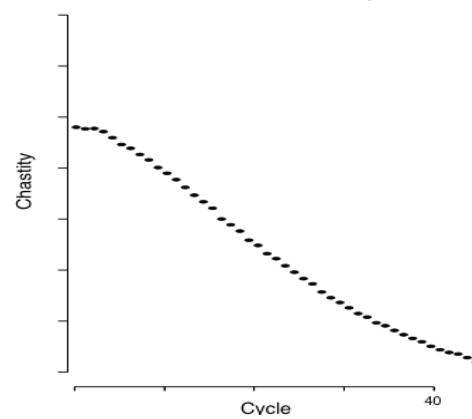


AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA
Noise removal

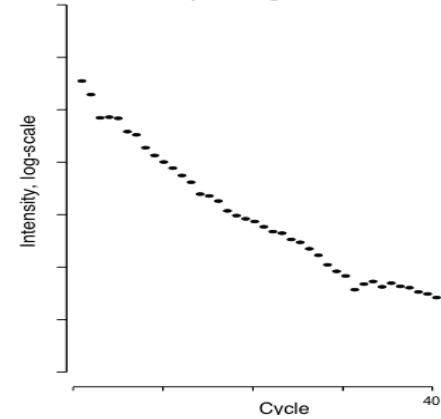
Raw data



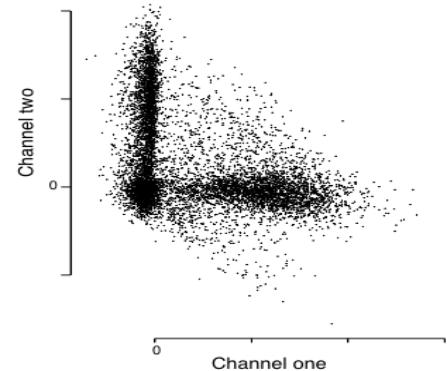
Median Chastity



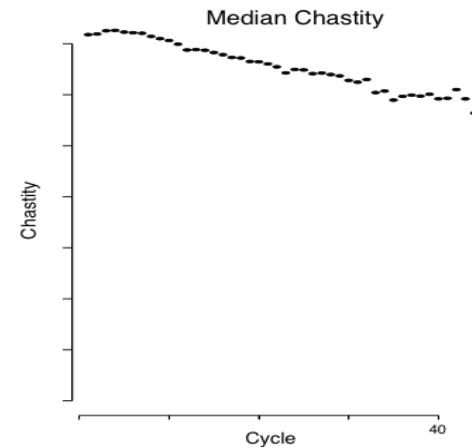
Intensity of Brightest Channel



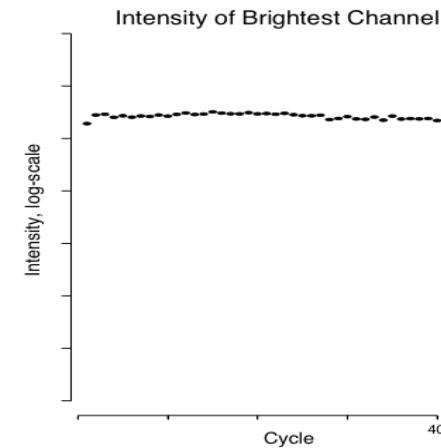
AYB processed



Median Chastity

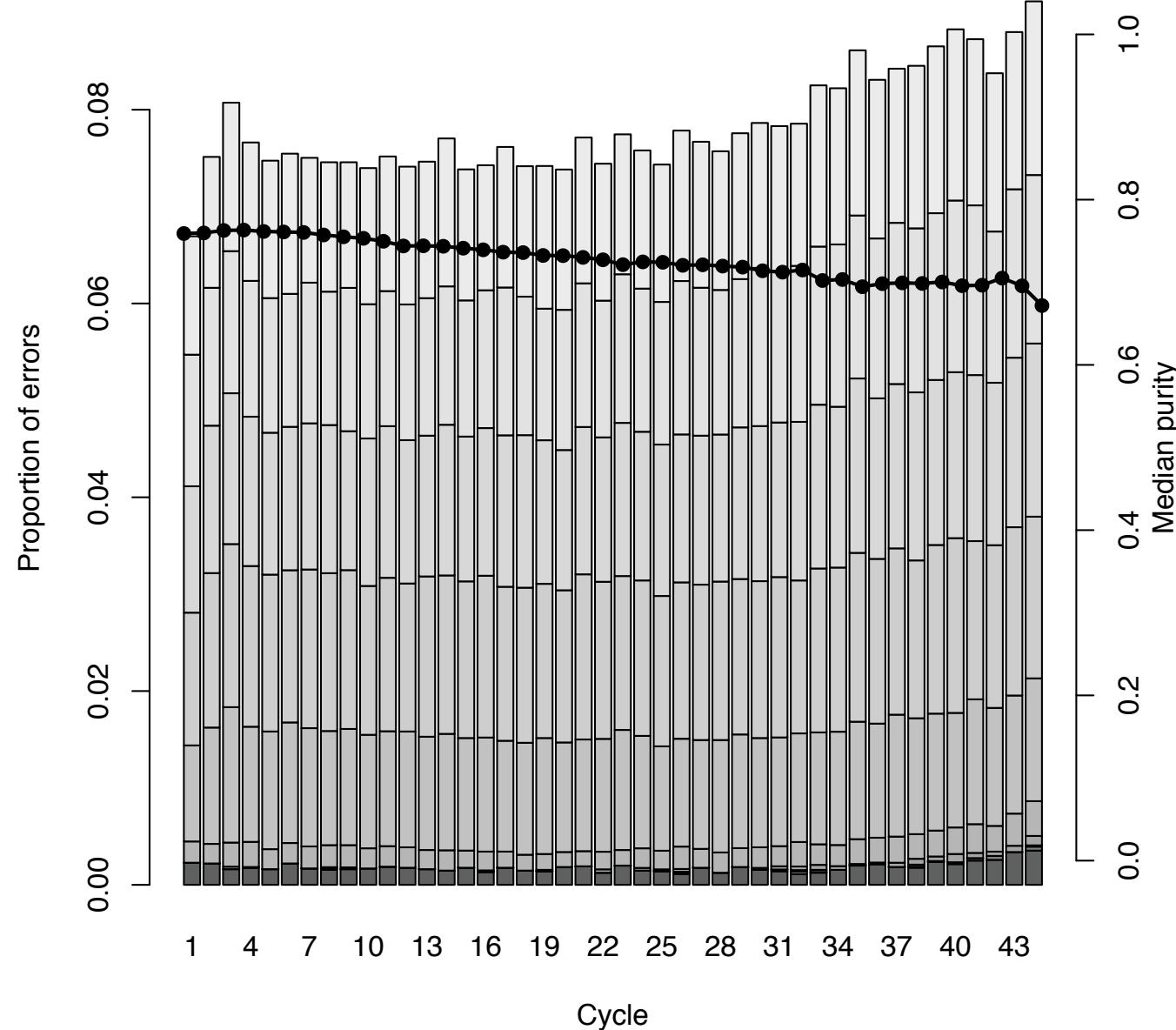


Intensity of Brightest Channel



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAA~~T~~GCCCAGA
Accuracy

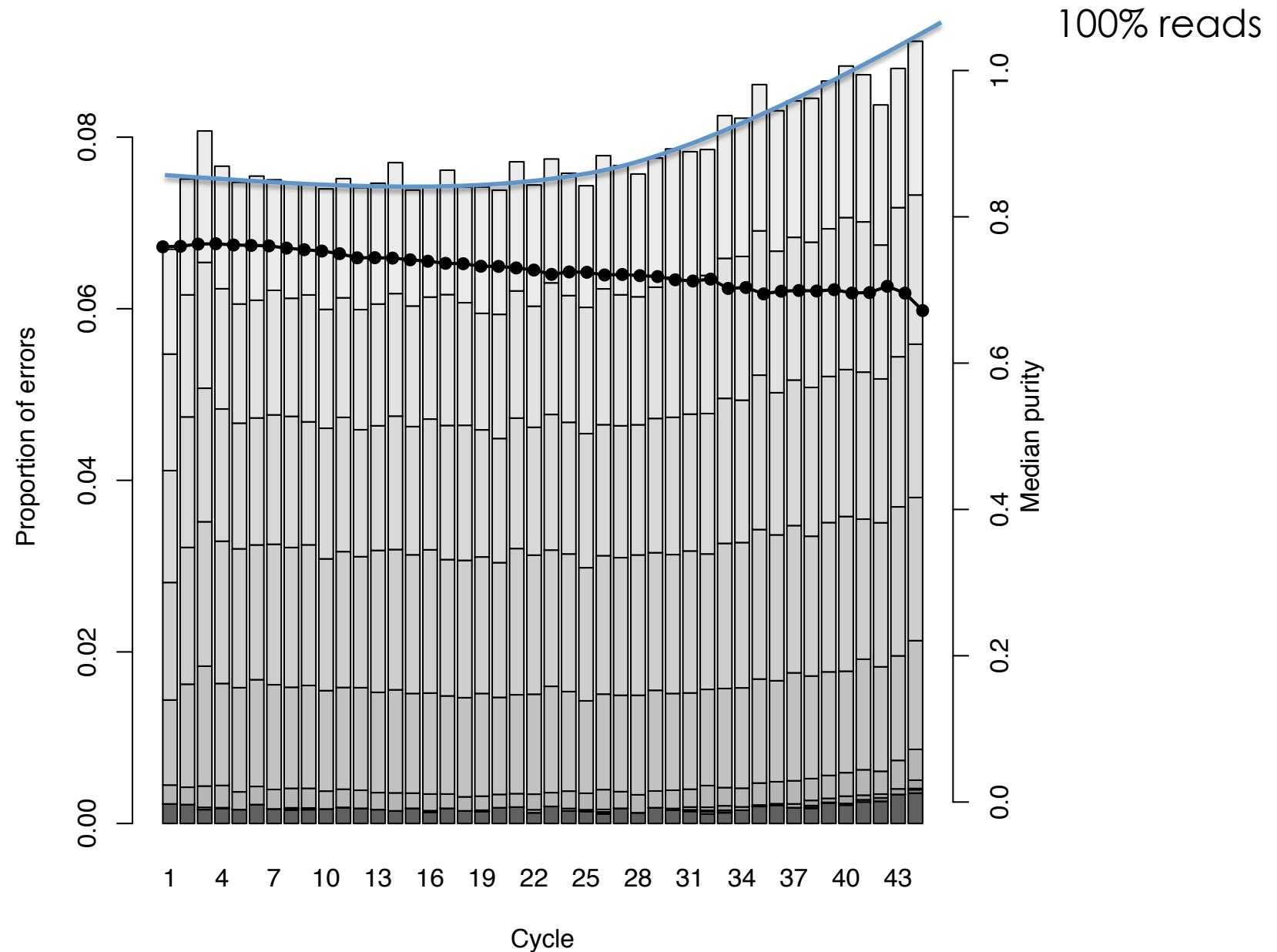
134358/135856 reads mapped (98.9 %)



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Accuracy

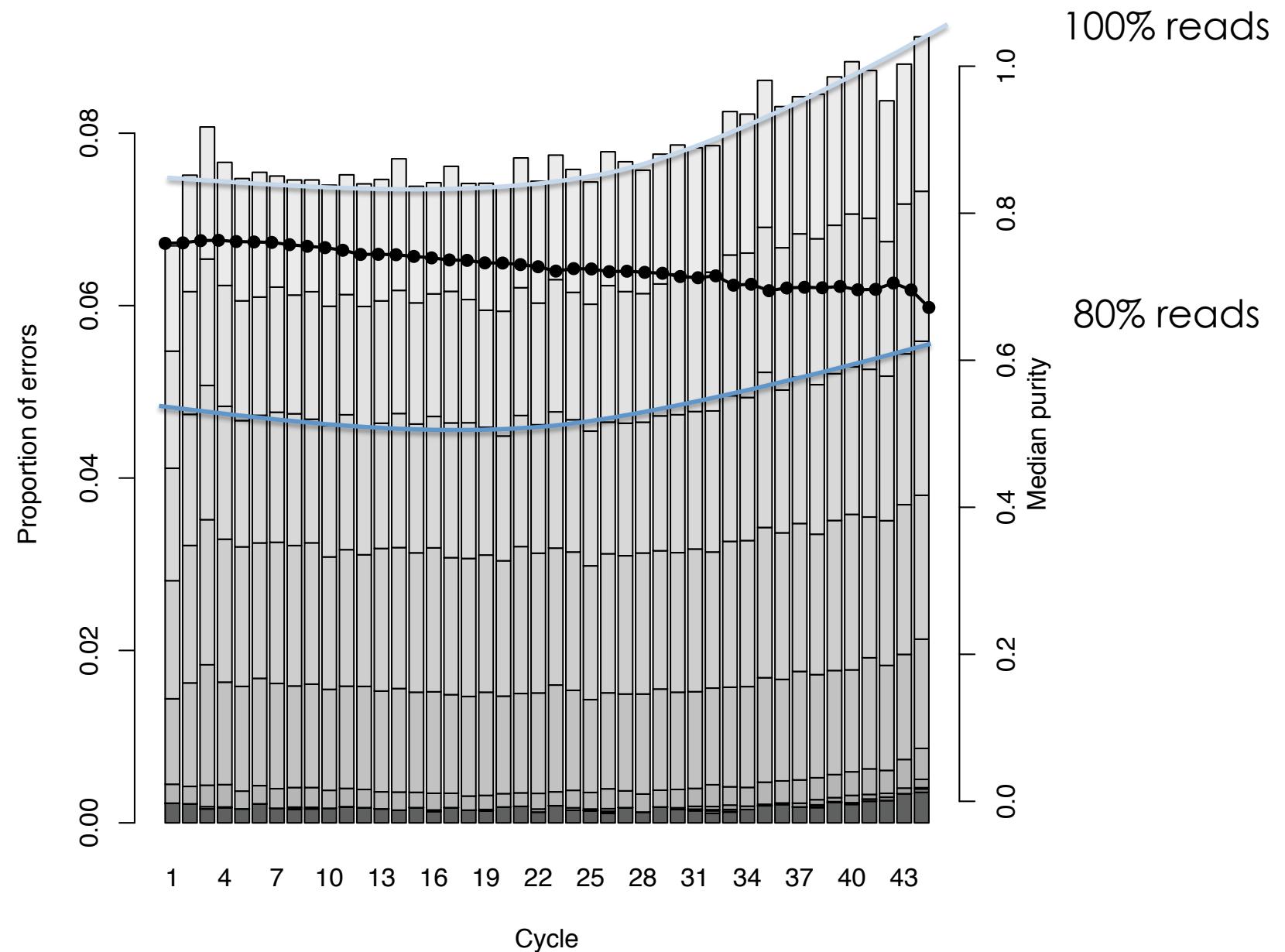
134358/135856 reads mapped (98.9 %)



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Accuracy

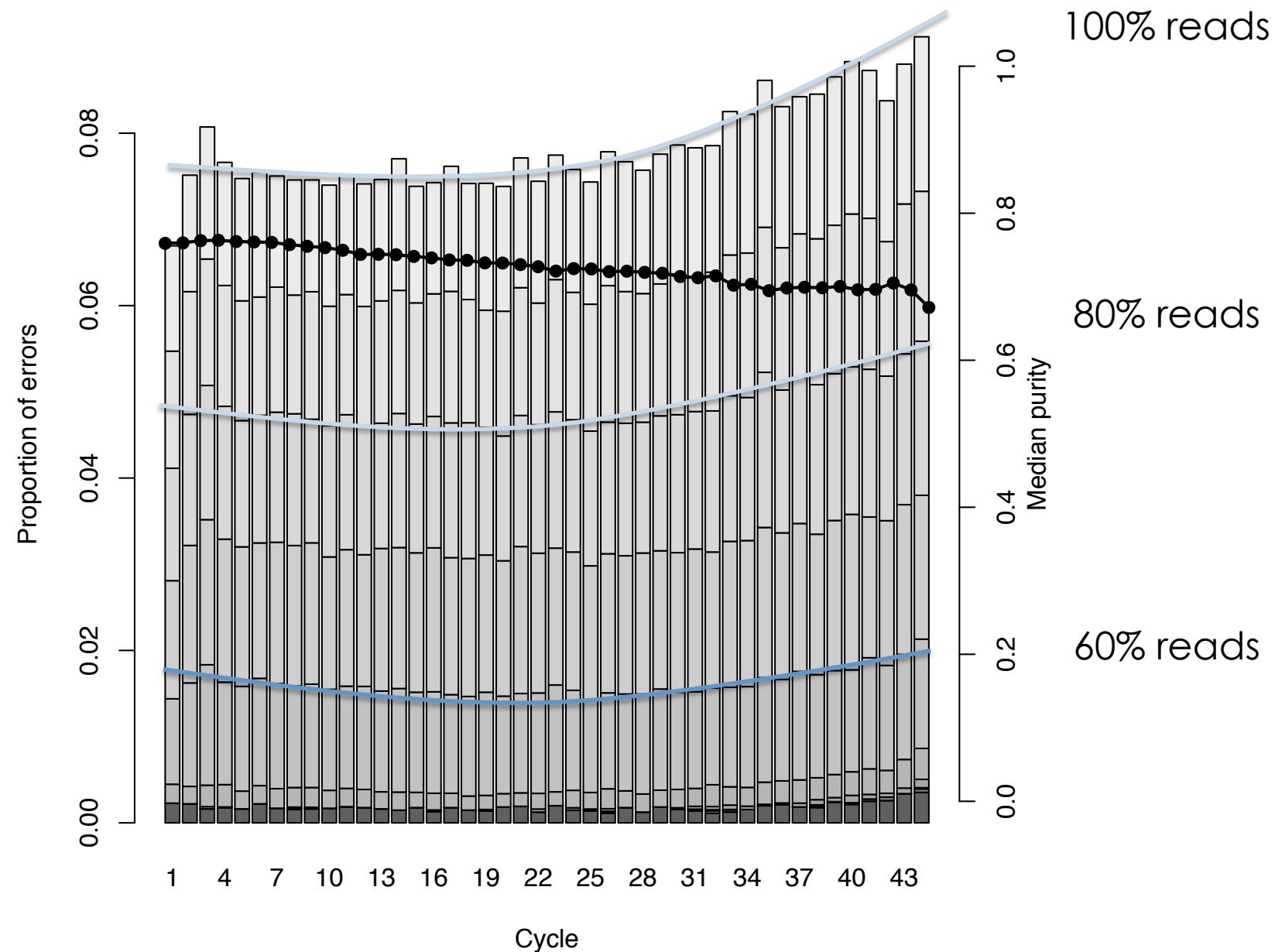
134358/135856 reads mapped (98.9 %)



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

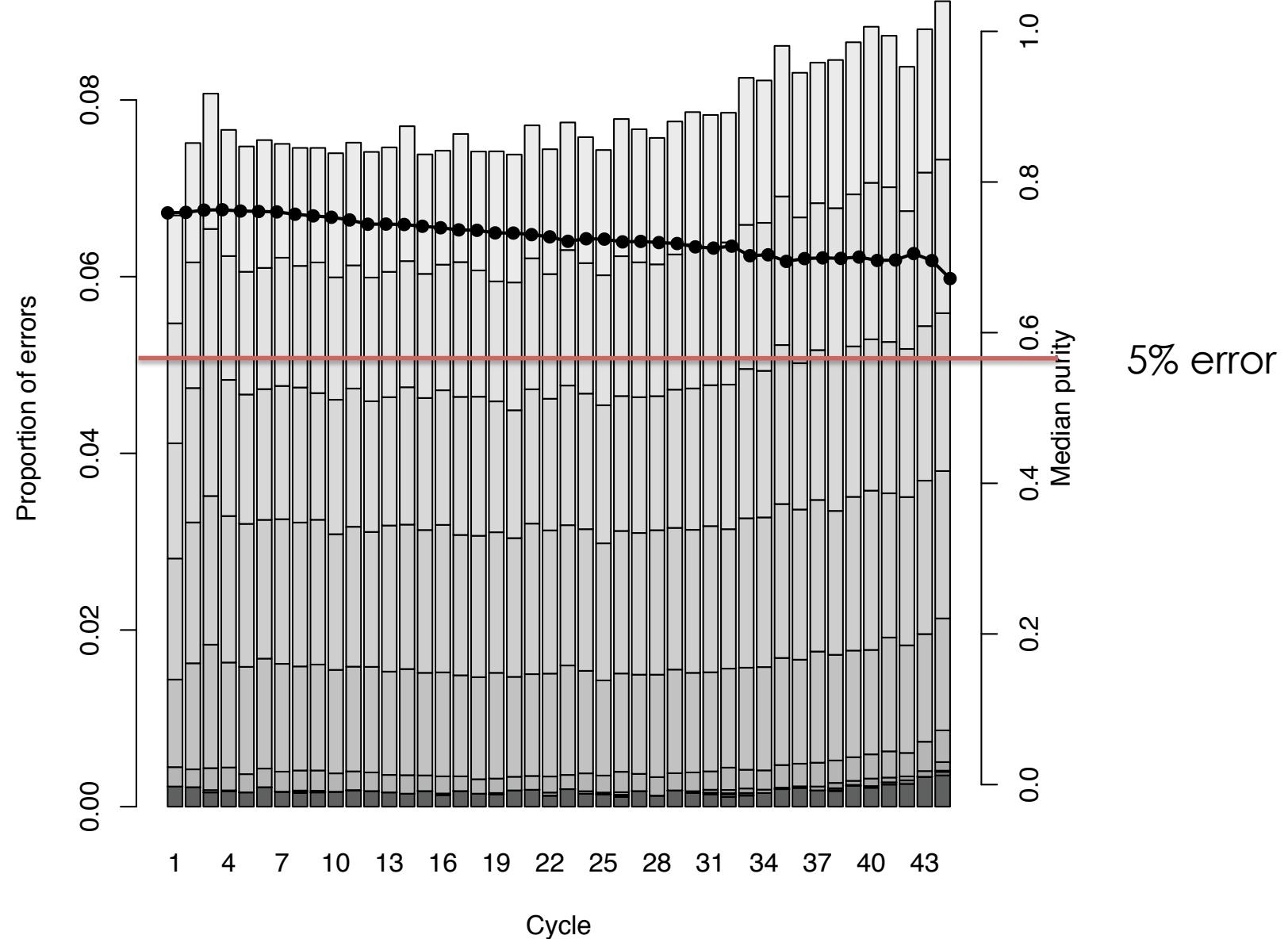
Accuracy

134358/135856 reads mapped (98.9 %)



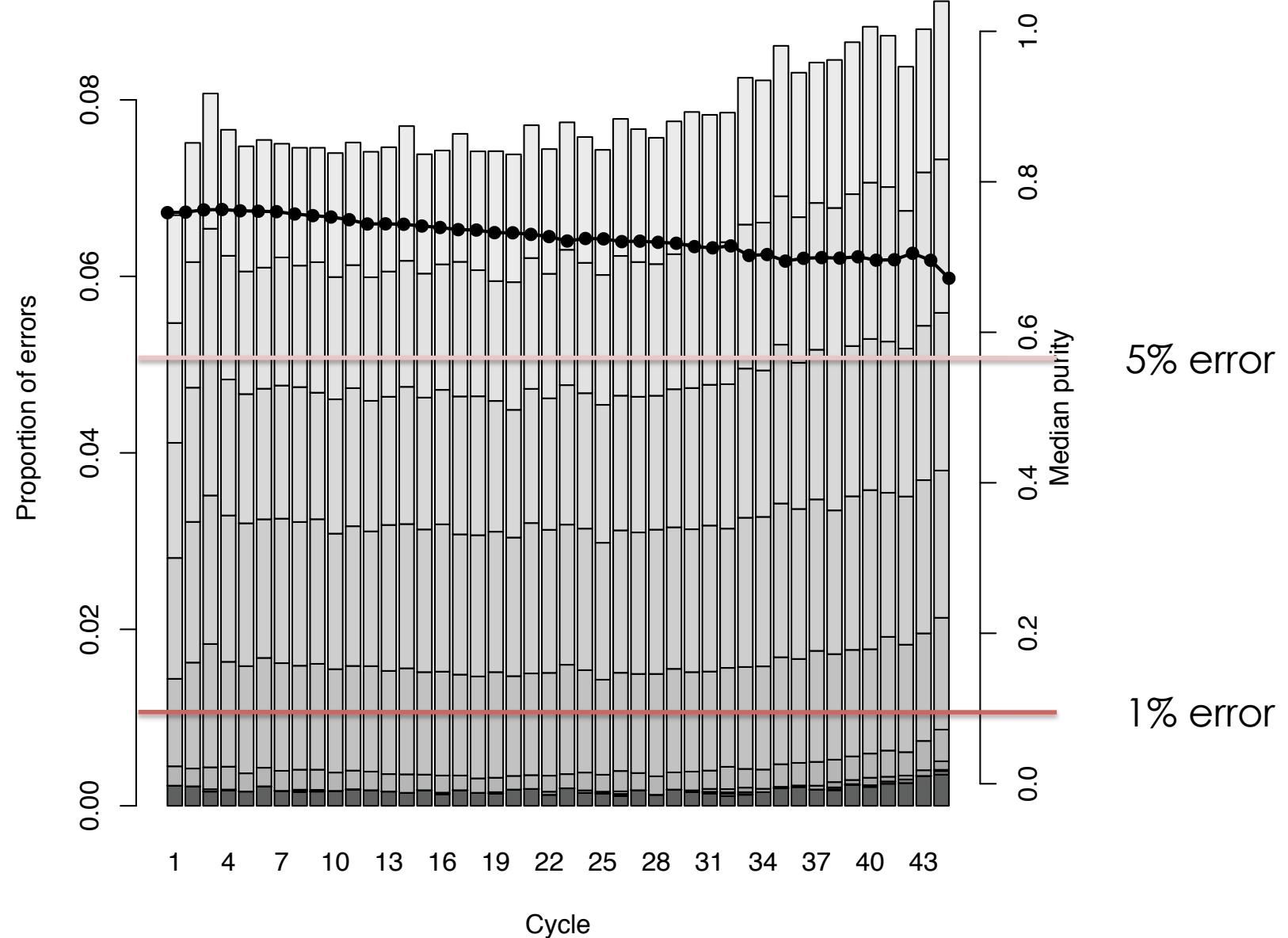
AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAA~~T~~GCCCAGA
Accuracy

134358/135856 reads mapped (98.9 %)



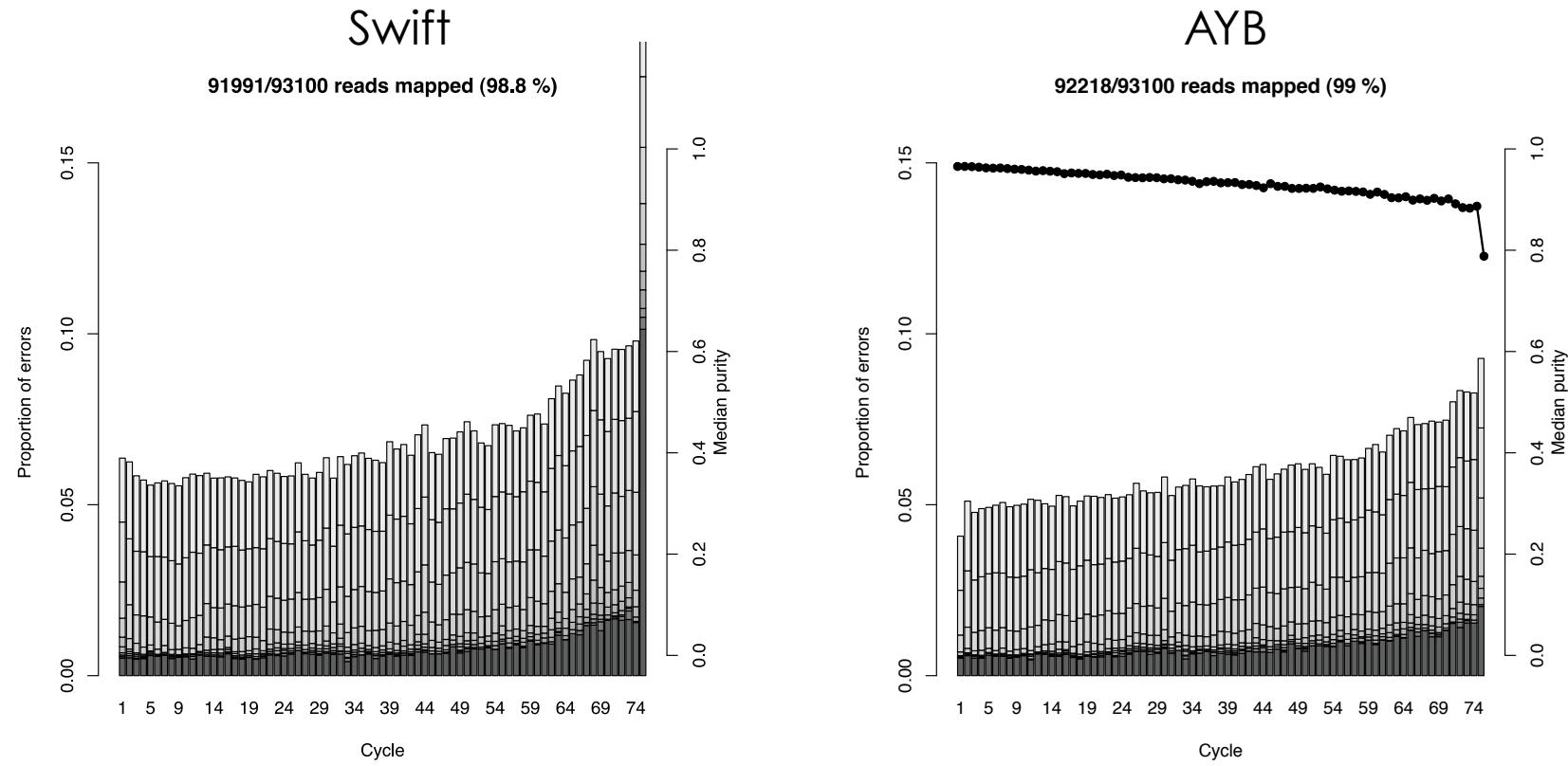
AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAA~~T~~GCCCAGA
Accuracy

134358/135856 reads mapped (98.9 %)



AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

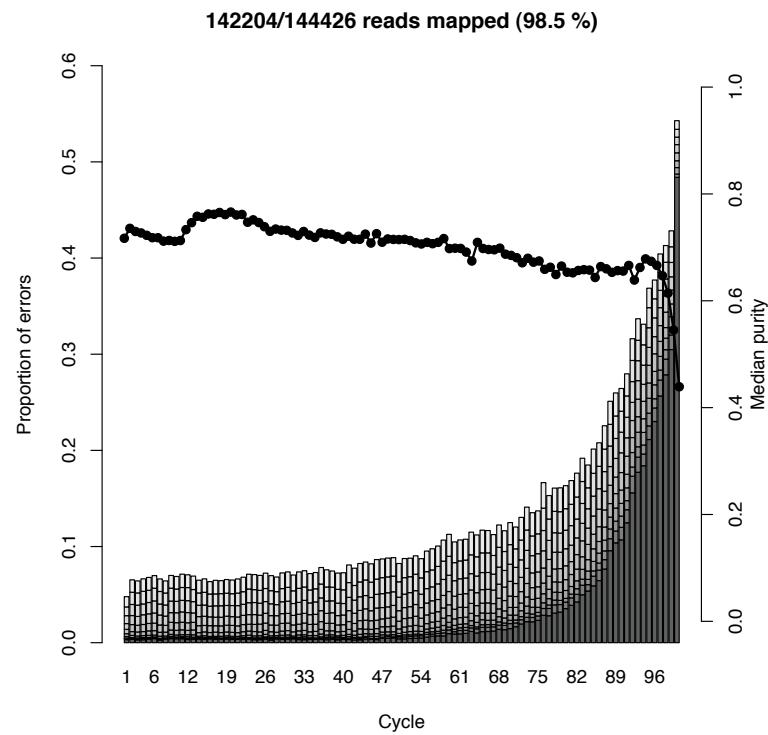
75 cycle comparison



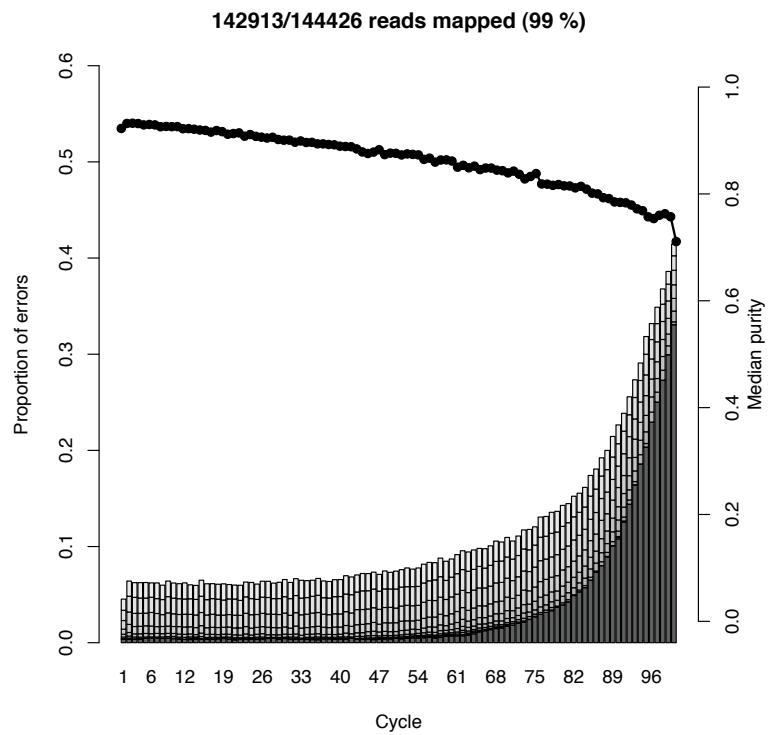
No purity values for Swift because of bug, ordered by purity for AYB

AGATAGGAAGAGCGGTCAGCACCGAATGCCGAGA
100 cycle accuracy

Swift

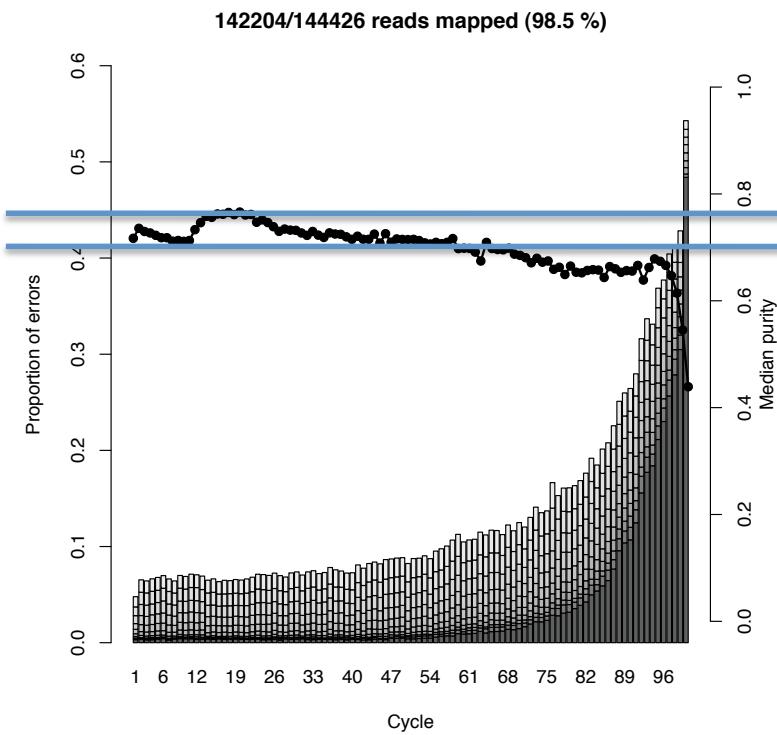


AYB

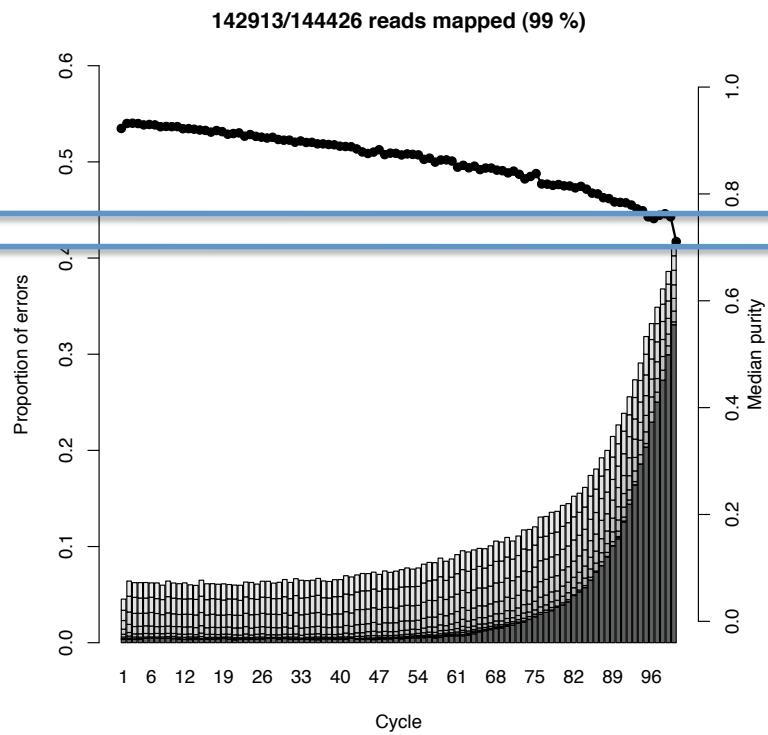


AGATAGGAAGAGCGGTCAGCACCGAATGCCGAGA
100 cycle accuracy

Swift



AYB

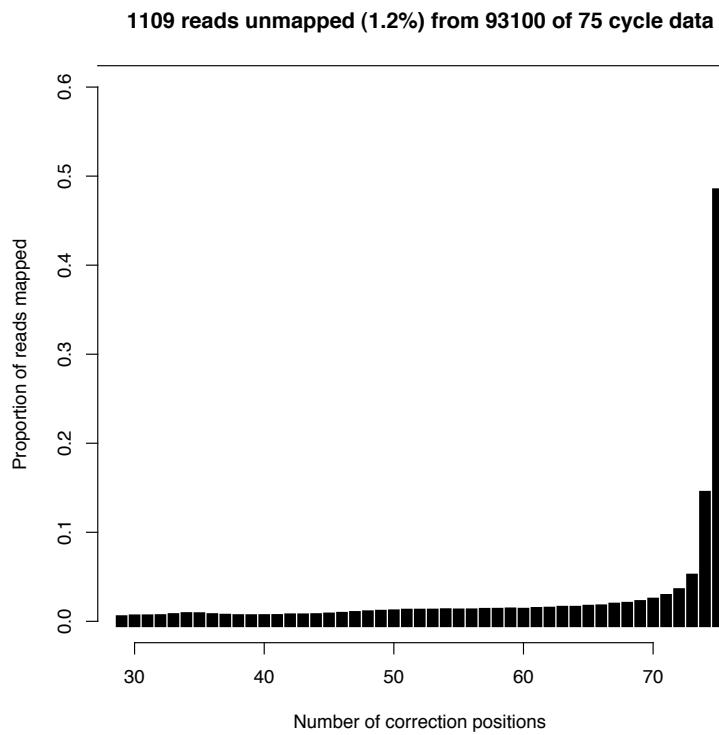


AGATAGGAAGAGCGGTCACGCCACGGATGCCGAGA

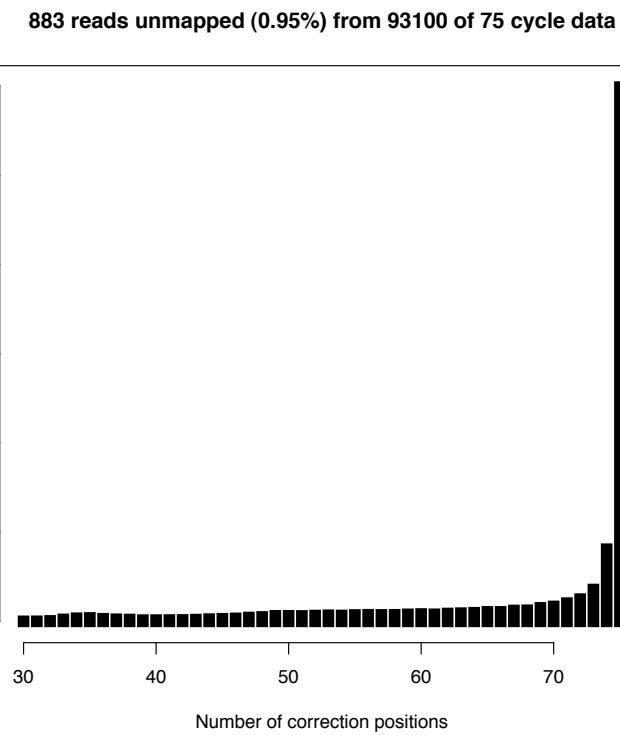
Number of correct reads

75 cycle phiX data

Swift



AYB



AYB improvement (BWA alignments, edit distance 7)

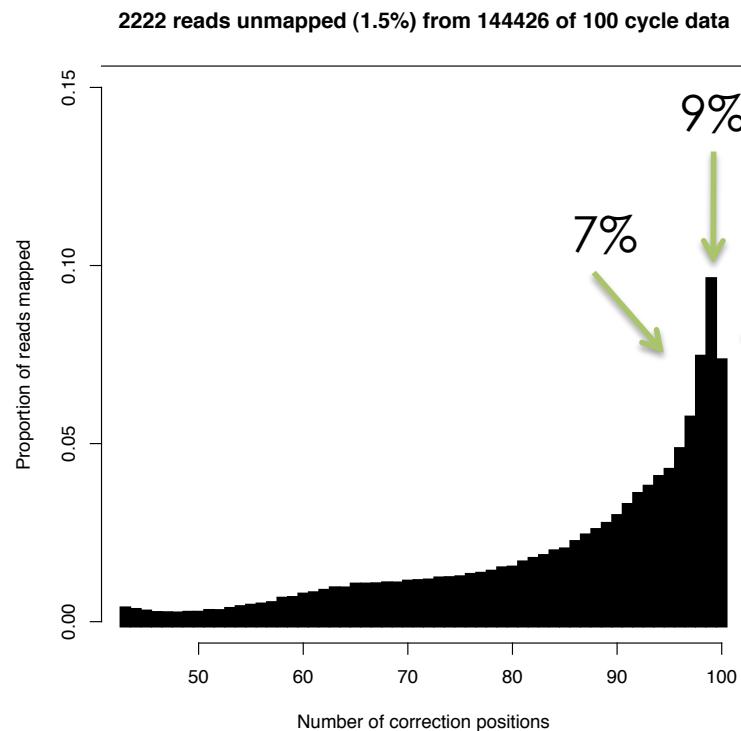
- 4% more reads aligned
- 25% more perfect reads

AGATAGGAAGAGCGGTCACGCCACGGATGCCGAGA

Number of correct reads

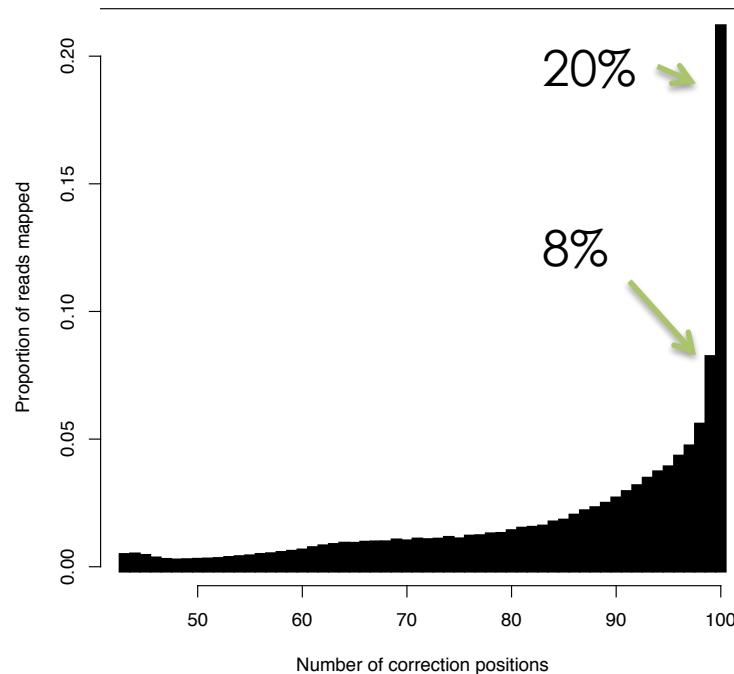
100 cycle phiX data

Swift



AYB

1540 reads unmapped (1.1%) from 144426 of 100 cycle data



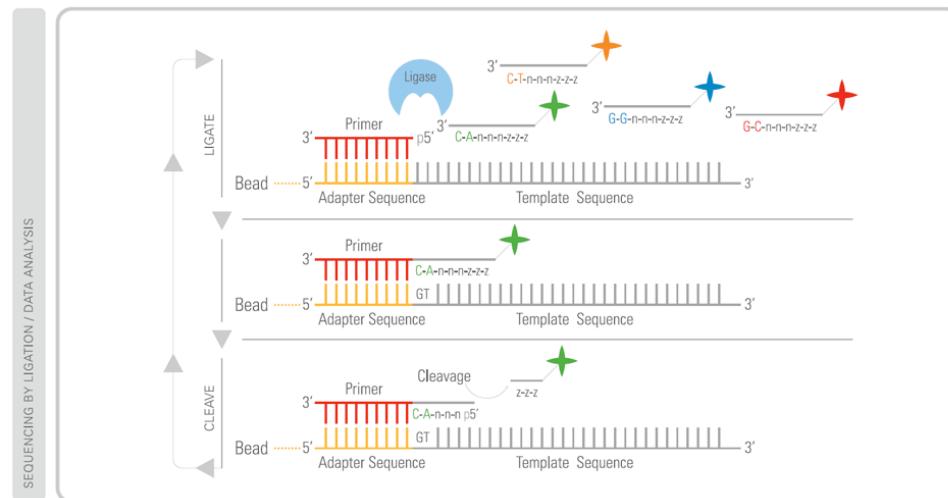
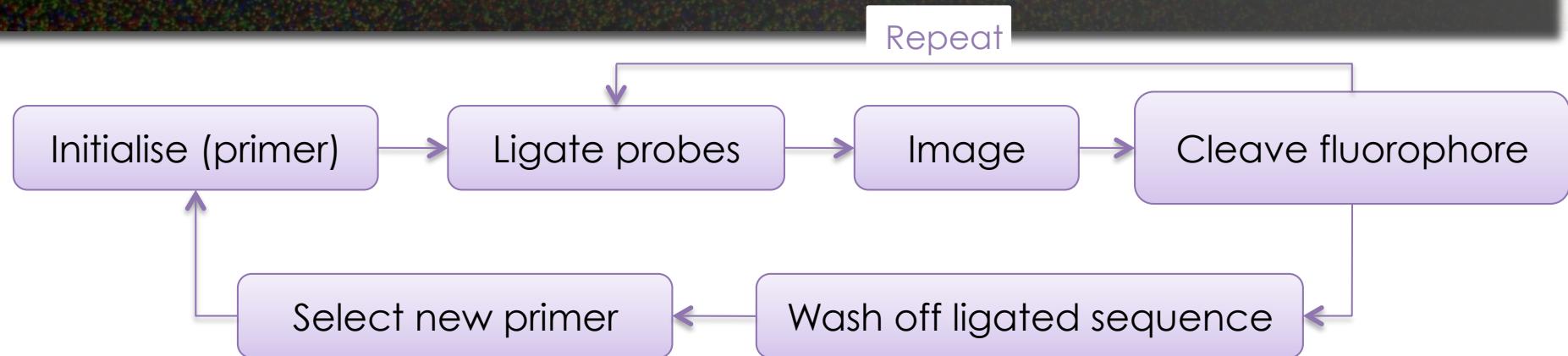
AYB improvement (BWA alignments, edit distance 7)

- 15% more reads aligned
- 180% more perfect reads

Will show later that ~25% of reads in this data set have contamination at final cycle

AGATAGGAAAGAGCGGTTCAGCAGGAATGCCGAGA

SOLiD chemistry

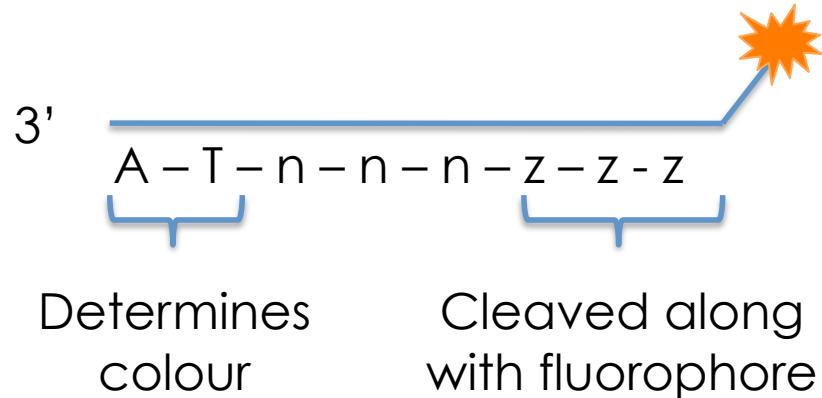


Additional step attaches permanent blockers to reads that were not extended

SOLiD trades phasing for dimming

- less opportunity for correction
- may still be able to improve calls

AGATAGGAAGAGCGGTTCA
GCCACGGAAATGCCGAGA
SOLiD probes



Errors

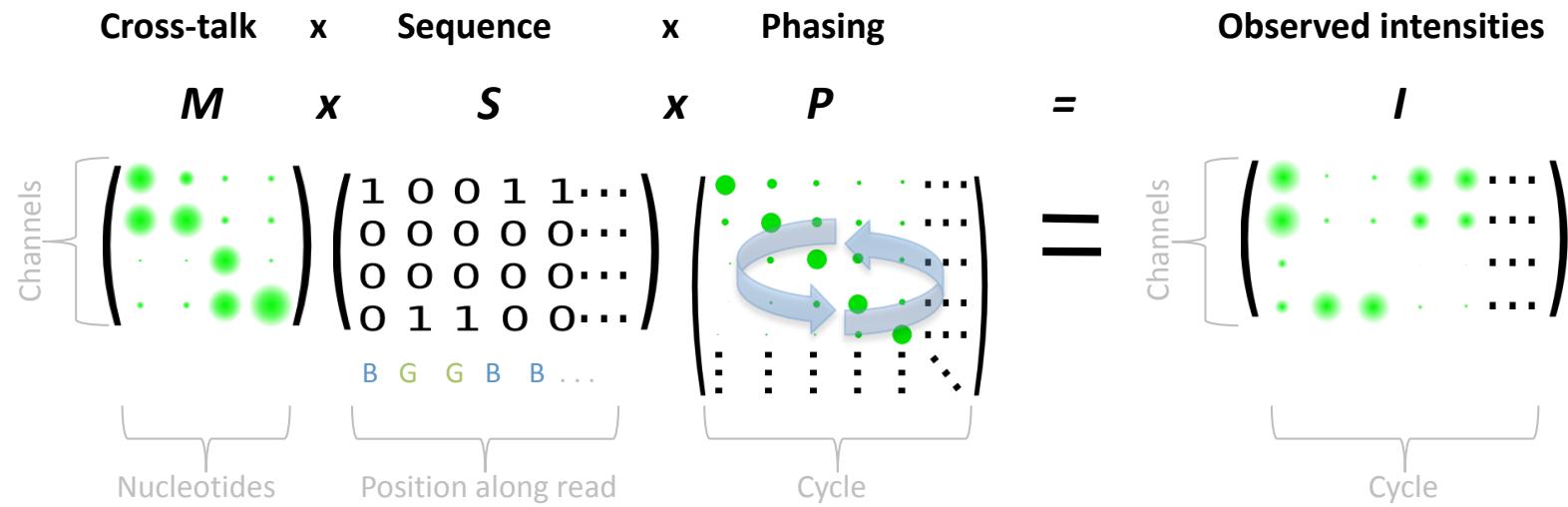
- Failure to ligate probe sequence
- Ligate multiple probe sequences (blocker failure)
- Incorrect cleavage (stop sequencing entirely, or leave additional bases)

All these are captured by phasing matrix

- Bad incorporation (similar but incorrect probe)
 - only problem if first two bases affected
 - transient error

AGATAGGAAGAGCGGTTCAGGAAATGCCGAGA
AYB on SOLiD

Same model as before, using colours rather than bases



Permutation of read position to cycle

- permute phasing matrix
- Primer reset
- intensity increases each primer round? Less dimming

All these covered since phasing is estimated empirically

AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

Calibration

Measure confidence in each base call

AYB - fit of each possible call to model $\mathbf{P}(\text{data} \mid \text{base is } A)$

Use robustified Bayesian approach

$$\mathbf{P}(\text{base is } A) = \frac{\mathbf{P}(\text{data} \mid \text{base is } A) + \eta}{\sum_{j \in \{A,C,G,T\}} \mathbf{P}(\text{data} \mid \text{base is } j) + \eta}$$

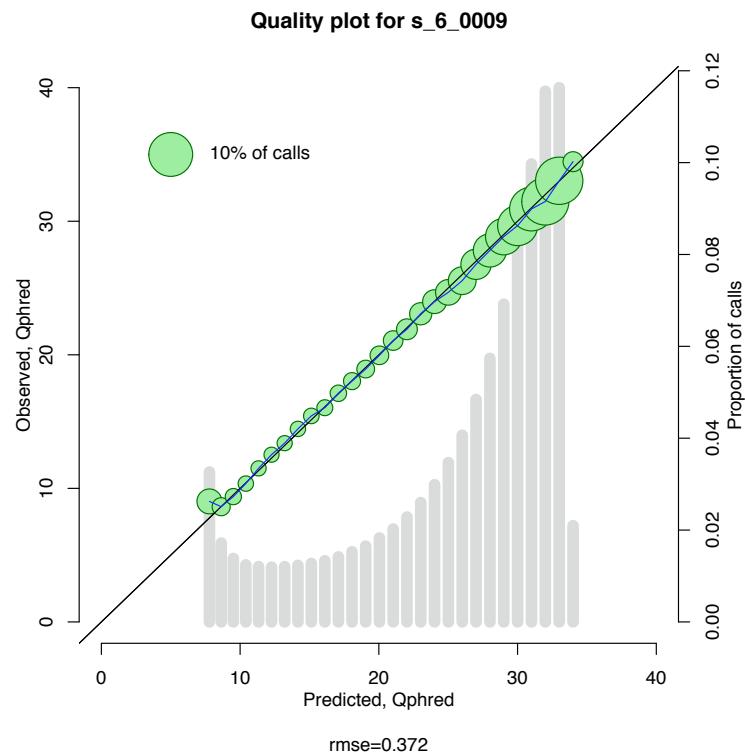
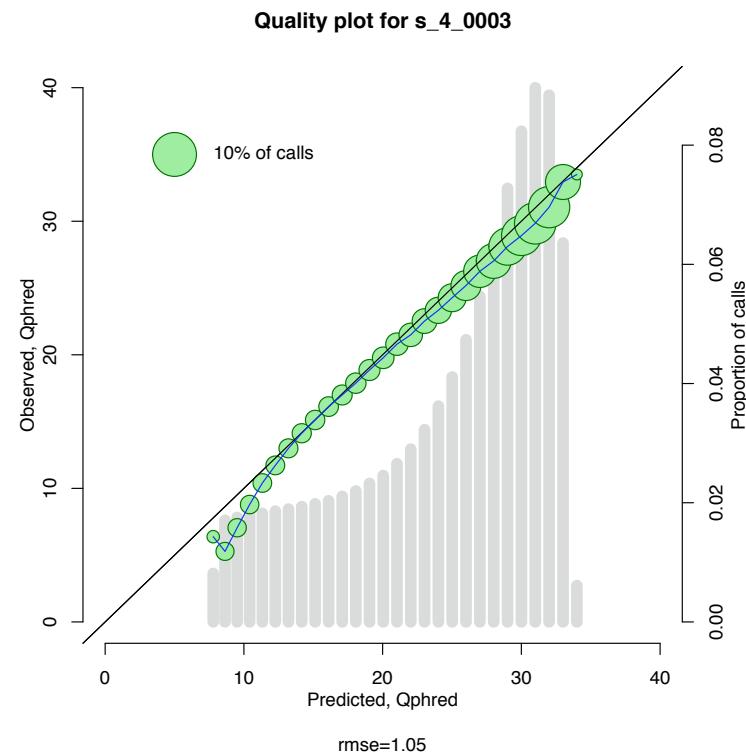
η represents “contamination” from other sources

$\eta \sim Q50$ by default

If none of the bases fit well, then posterior probability tends to 0.25
Confidence resets when data does not look like sequence

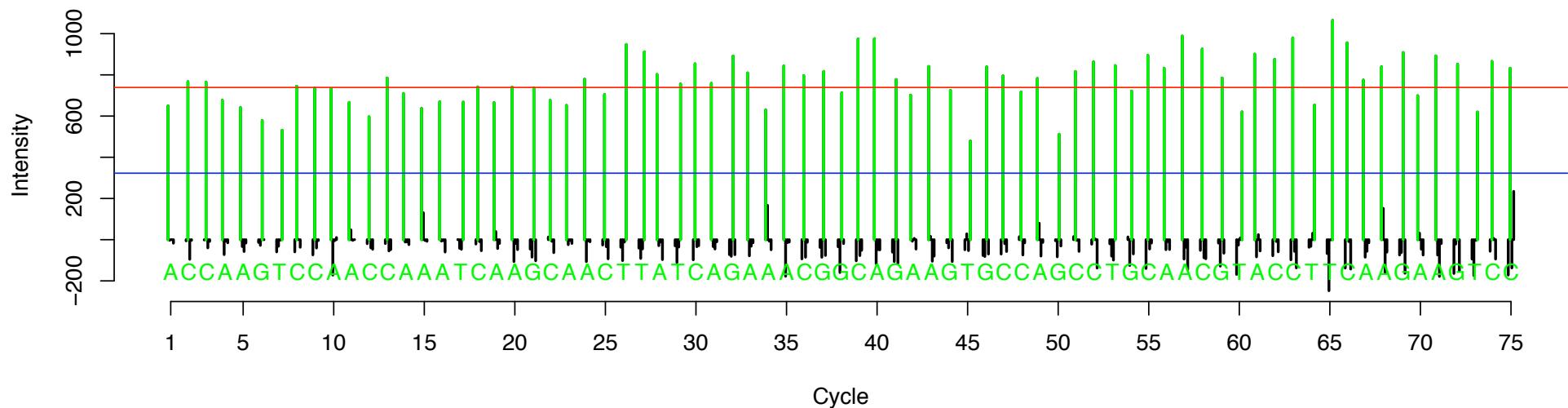
AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA

Calibration



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Assessing quality



Unmodelled effects are upper bound on quality

From:

$$P(B \cup O) = P(B) + P(O) - P(B \cap O) \leftarrow \in [0, \min P(B), P(O)]$$

Total error Other effects
Base calling error

Get:

$$\min(Q_{\text{Base}}, Q_{\text{Effect}}) \geq Q_{\text{Total}} \geq \min(Q_{\text{Base}}, Q_{\text{Effect}}) - 3$$

E.g. polymerase error ~ Q40

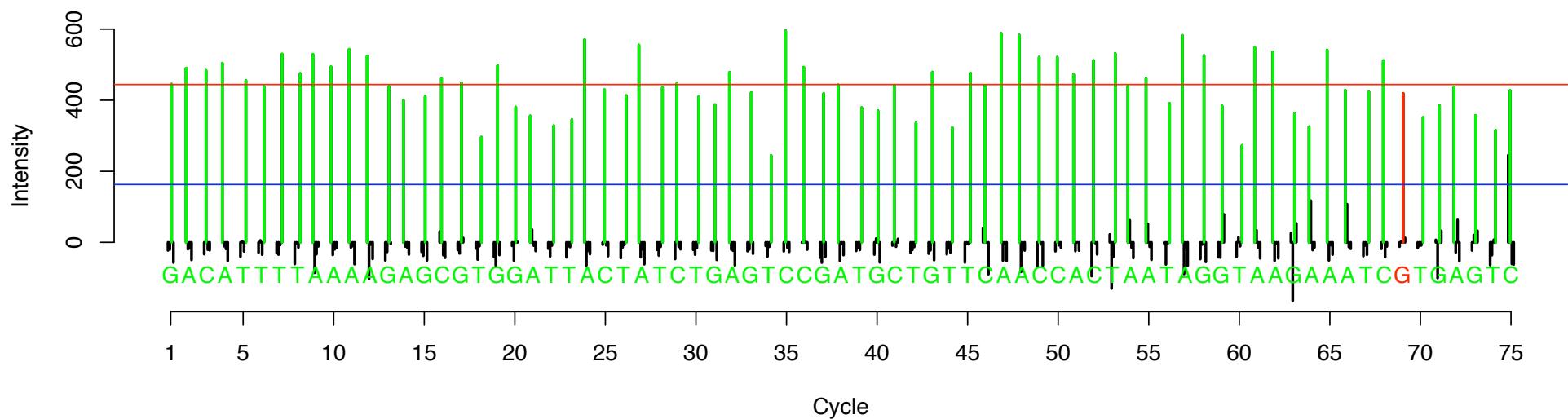
AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Early polymerase errors

Polymerase error during sample preparation or early amplification

Indistinguishable from a SNP

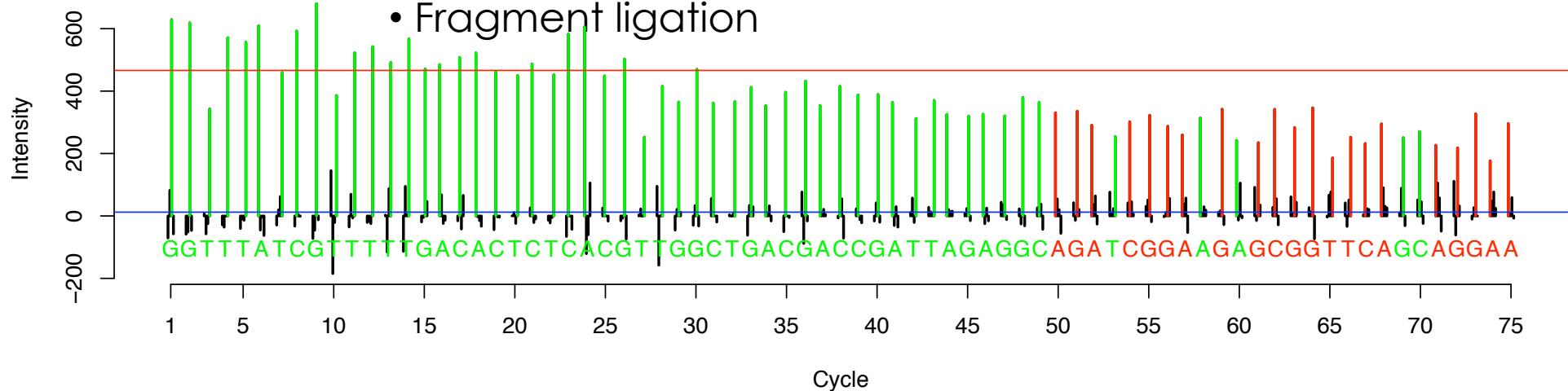
- sequence with a difference from reference
- call sequence in cluster correctly but get error



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA
Errors in read

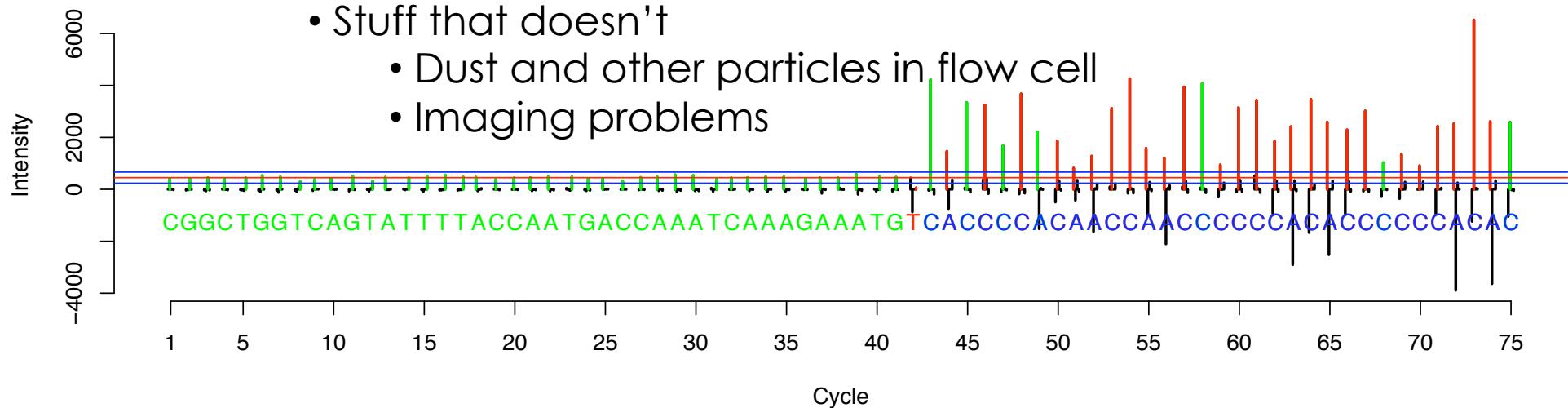
- Stuff that looks like sequence

- E. coli, H. sapiens etc sequence contamination
- Bits of replication machinery
- Adapter sequence
- Fragment ligation



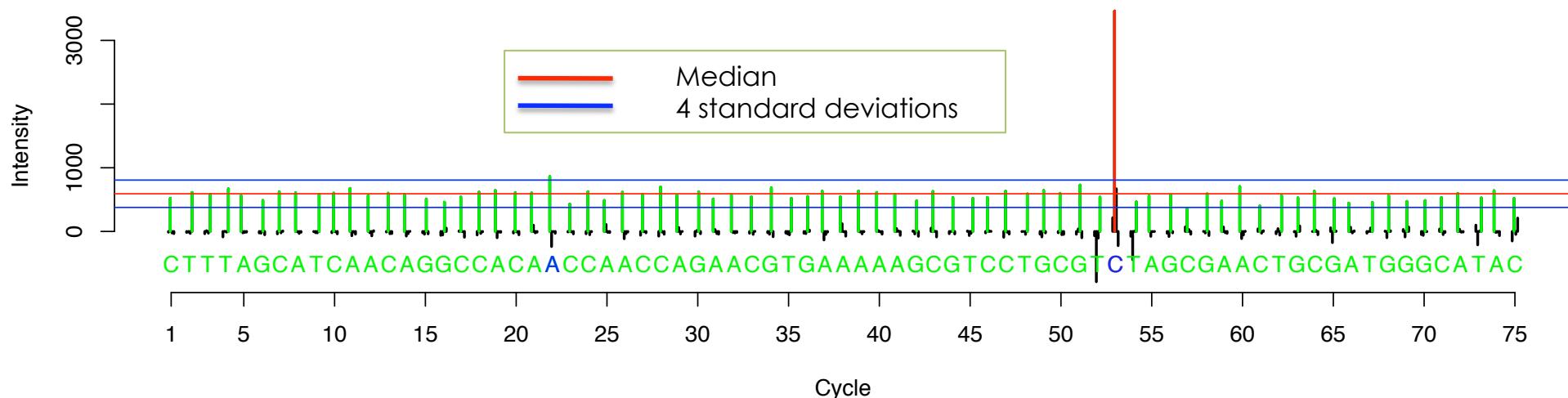
- Stuff that doesn't

- Dust and other particles in flow cell
- Imaging problems



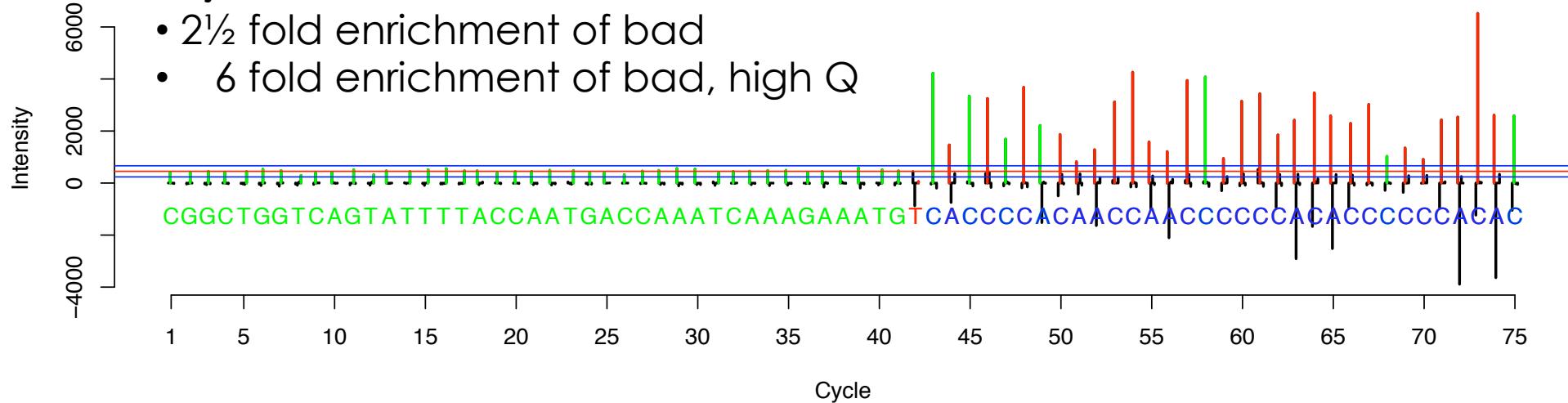
AGATAGGAAGAGGCCGTTCAGCAGGAATGCCGAGA Filtering artefacts

Observation: contaminants are much brighter than ordinary sequence



Reject 0.8% of bases

- 2½ fold enrichment of bad
 - 6 fold enrichment of bad, high Q



AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

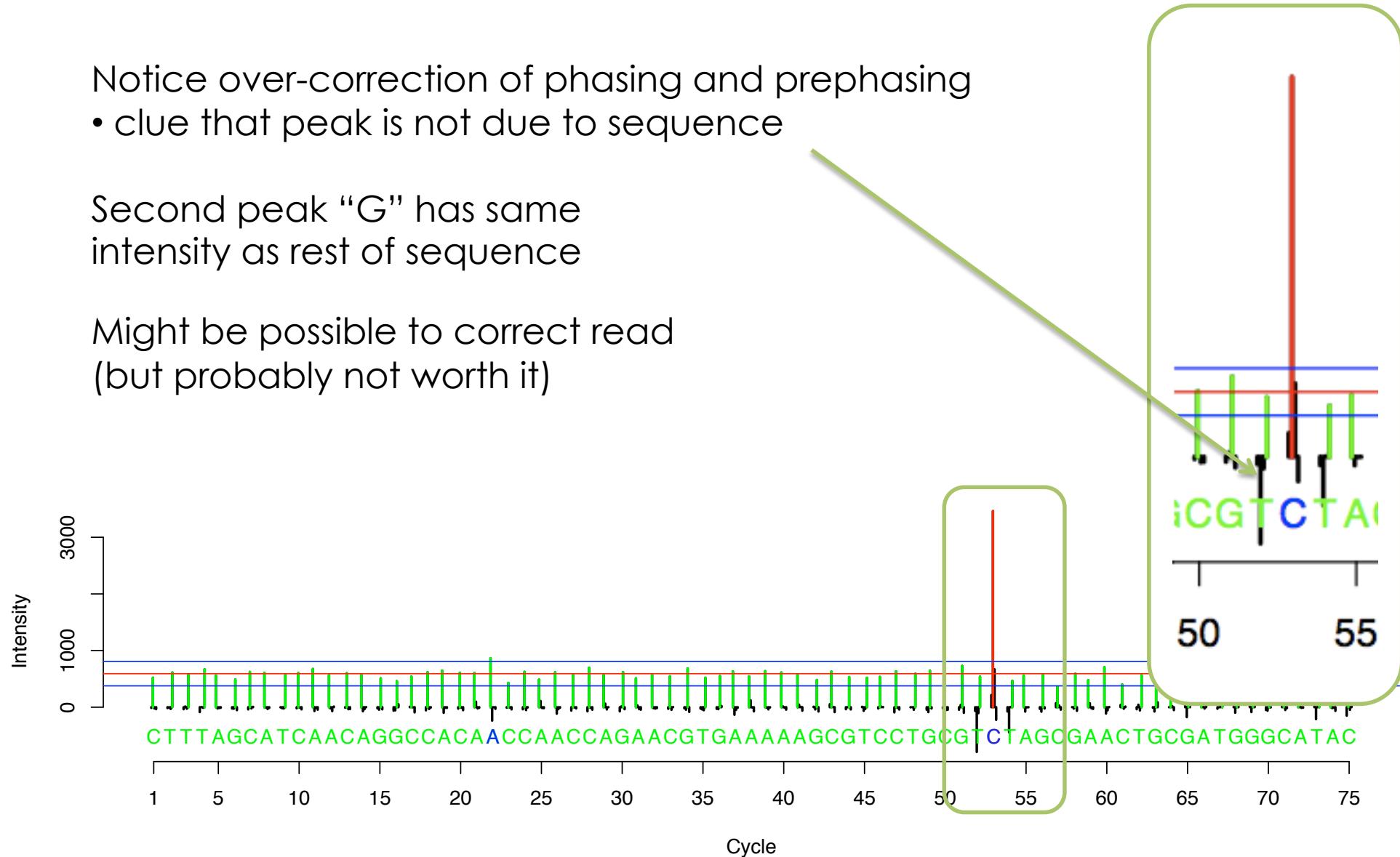
Filtering artefacts

Notice over-correction of phasing and prephasing

- clue that peak is not due to sequence

Second peak "G" has same intensity as rest of sequence

Might be possible to correct read
(but probably not worth it)

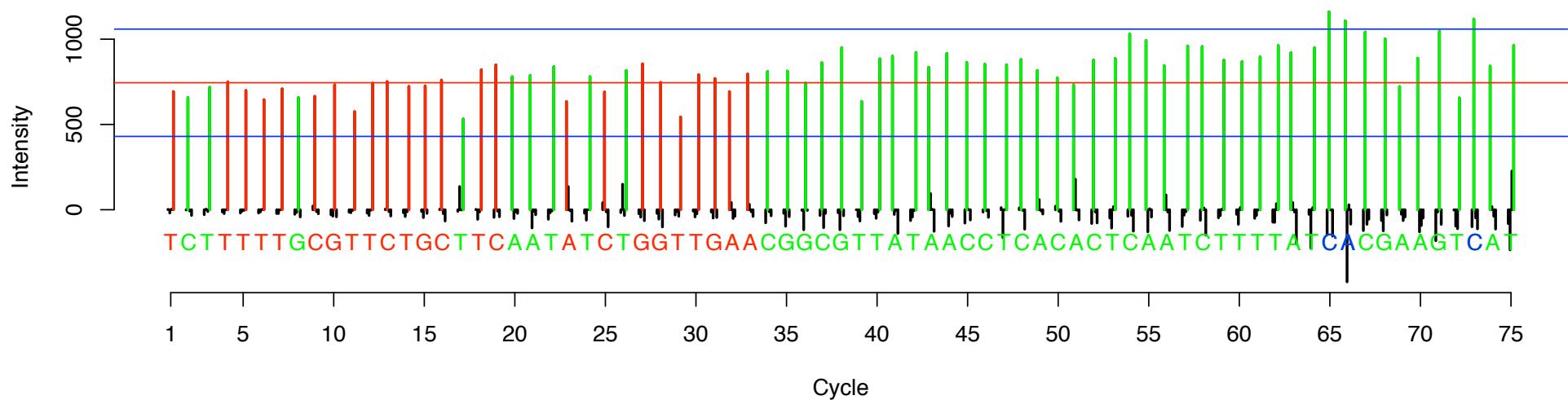


AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Fragment ligation

Two fragments of DNA can ligate before sequencing

- Apparently good read
- High error rate
- Rare



TCTTTTGCCTCTGCTTCAATATCTGGTTAACGGCGTCGCGTCGTAAACCCAGCTTGGTAAGTTGGATTAAAGCA PhiX 5190 -ve
|||||||

TCTTTTGCCTCTGCTTCAATATCTGGTTAACGGCGTTATAACCTCACACTCAATCTTTATCACGAAGTCAT Read

|||
CCTCAGCGGAAAAATTAAAATTACCGCTTCGGCGTTATAACCTCACACTCAATCTTTATCACGAAGTCAT PhiX 2273 -ve

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Polymerase slippage

20 base repeat

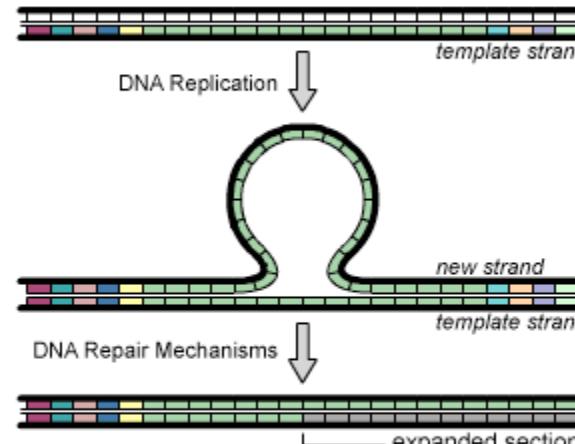
The diagram shows a DNA sequence with a bracket above it labeled "20 base repeat". Below the sequence, two lines of text represent reads from "phi - X174":
Read 1 (positions 1 to 34): CGTCACGTTATGGTGAACAGTGGATTAAGTTCA
Read 2 (positions 35 to 75): CGTCACGTTATGGTGAACAGTGGATTAAGTTCA
 |
 |
 TGAACAGTGGATTAAGTTCA
 |
 |
 TGAAGGGATGGTGTAAATGCCA

Templated sequence

Polymerase slips during replication causing a region to be repeated

Figure Q-5: The Polymerase Slippage Model

A) Slippage Event



(A) During replication, polymerase slippage and subsequent reattachment may cause a bubble to form in the new strand. Slippage is thought to occur in sections of DNA with repeated patterns of bases (such as CAG), represented here by matching colors. Then, DNA repair mechanisms realign the template strand with the new strand and the bubble is straightened out. The resulting double helix is thus expanded.

B) No Slippage

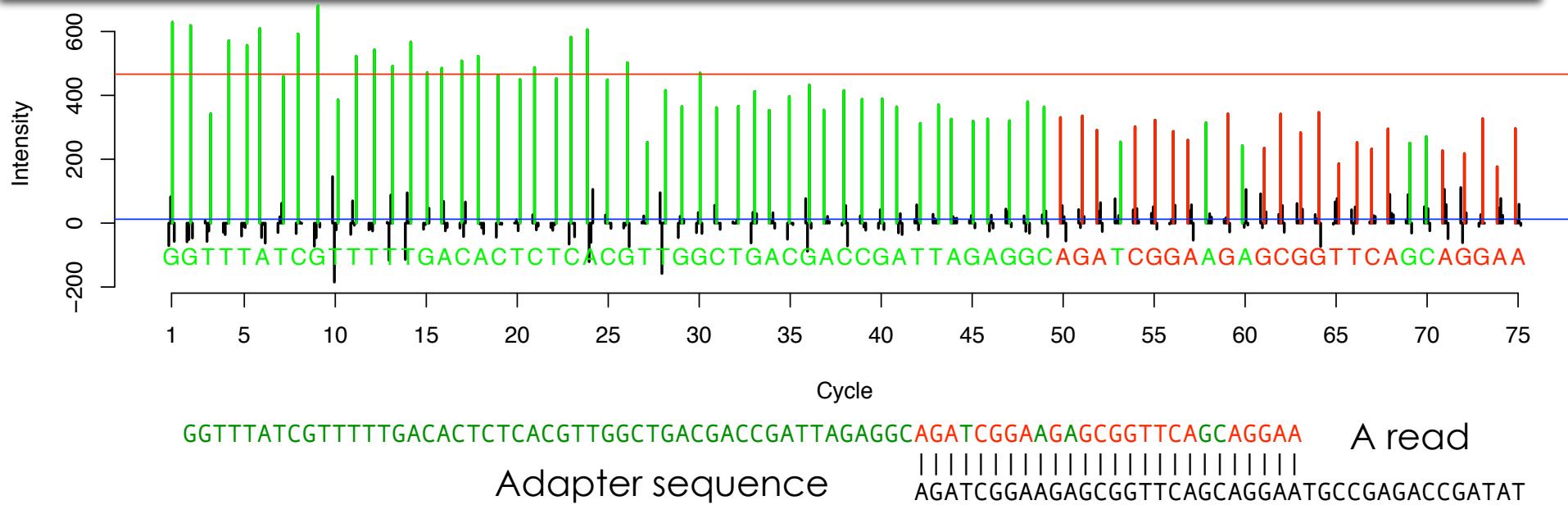


(B) Polymerase slippage, as theorized, cannot occur in DNA without repeating patterns of bases.

AGATAGGAAGAGCGGTTCAAGCAGGAATGCCGAGA

Adapter Sequence

AGATCGGAAGAGCGGTTCAAGCAGGAATGCCGAGACCGATAT



A couple of reads

TACCAATGACCAAATCAAAGAACGTTAGTCGCAAGGTTAGTGTGAGGTTGACTTAGATCGGAAGAGCGGTTCAGC
 TCATGAGTCAAGTTACTGAACAATCCGTACGTTCCAGACCGCTTGAGATCGGAAGAGCGGTTCAGCAGGAAATC
 AATATCAGCACCAACAGAAACAAACCTGATTAGCGGCGTTGACAGATGTATCCATCTGAAGATCGGAAGAGCGGTT
 TCAGAAAAGAGATTGCCGAGATGCAAAATGAGACTCAAAAGAGATTGCTGGAGATCGGAAGAGCGGTTCAGCAGG
 AGTAATCACGTTCTGGTCAATATAACCAGTAGTGTAAACAGTCGGGAAGATCGGAAGAGCGGTTCAGCAGGAAT
 TGACTATTCCAATGCAAAACTGAACGGCCTGGAAACACTGGTCATAATCATGGTGGCGAGATCGGAATGAGCGGT
 TGAGGTTATAACGCCGAAGCGGTAAGGTTAGATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGATAT
 AACCGAGAACGTGAAAAGCGCTCTGCGTAGCAGATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGATATC
 GATCGGAAGAGCGGTTAGCAGGAATGCCGAGATCGGAAGAGCGGTTCAGCAGGAATGCCGAGACCGATATCGTA
 ATTCAAGTCGGCGACTTCACGCCAGAACGAAAGACCAGGTATATGCACAAAAGATCGGAAGAGCGGTTCAGCAG

Note: Adapter sequence derived from study of Sanger Institute reads, yours may differ

AGATAGGAAGAGGCGGTTCAGCAGGAATGCCGAGA
A crude method to locate adapters

Search for reqd tails

- Starting with AGAT
 - >90% ID with adapter sequence
 - Length at least 8 bases

```

Best ungapped hit of adapter to phiX
PhiX      AGAACGAGAAGACGGTTACGCAGTTTGCCT
                  ||||.||||..|...||||||..|||||.|||..|||||
Adapter   AGATCGGAAGAGCGGTTACGCAGGAATGCCG

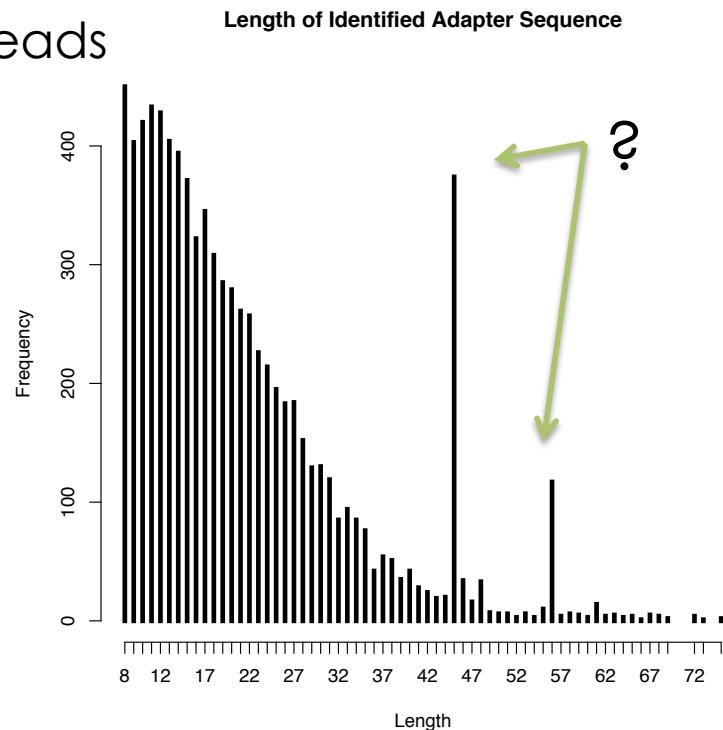
```

75 cycle data: about 0.08% of bases, 0.3% of reads

71% bases miscalled for adapter set
c.f. 6% bases miscalled for non-adapter set

Affect on quality

44 cycles	Q40	1 in 10,000
75 cycles	Q31	8 in 10,000
100 cycles	Q16	25 in 1,000



Average. Worse as cycle number increases

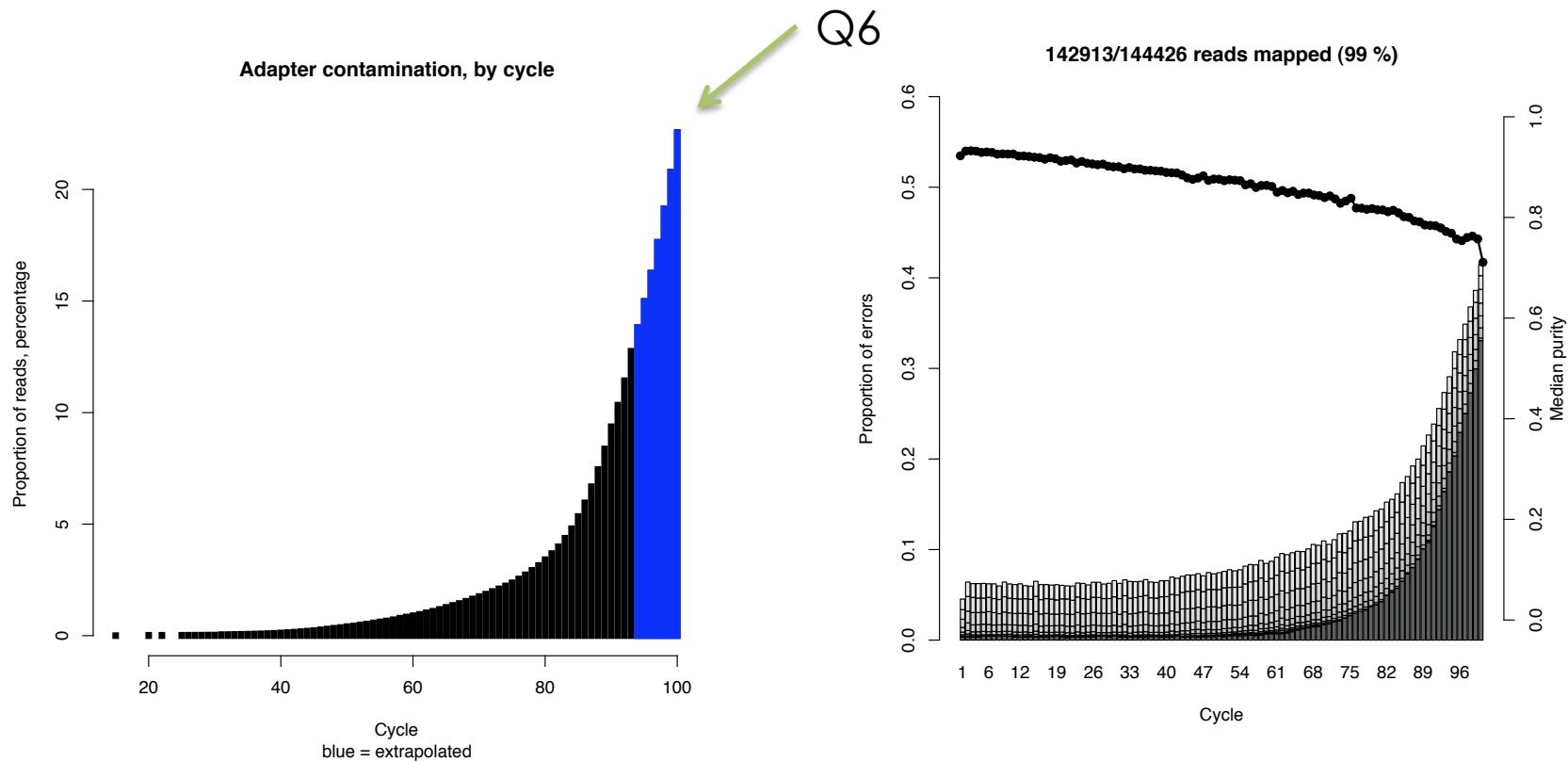
AGATAGGAAGAGCGGTCAGCAGGAATGCCGAGA 100 cycle data

Adapter contamination is accumulative

- starts rare but total effect can be large

Extrapolate number of adapters to final 7 sites

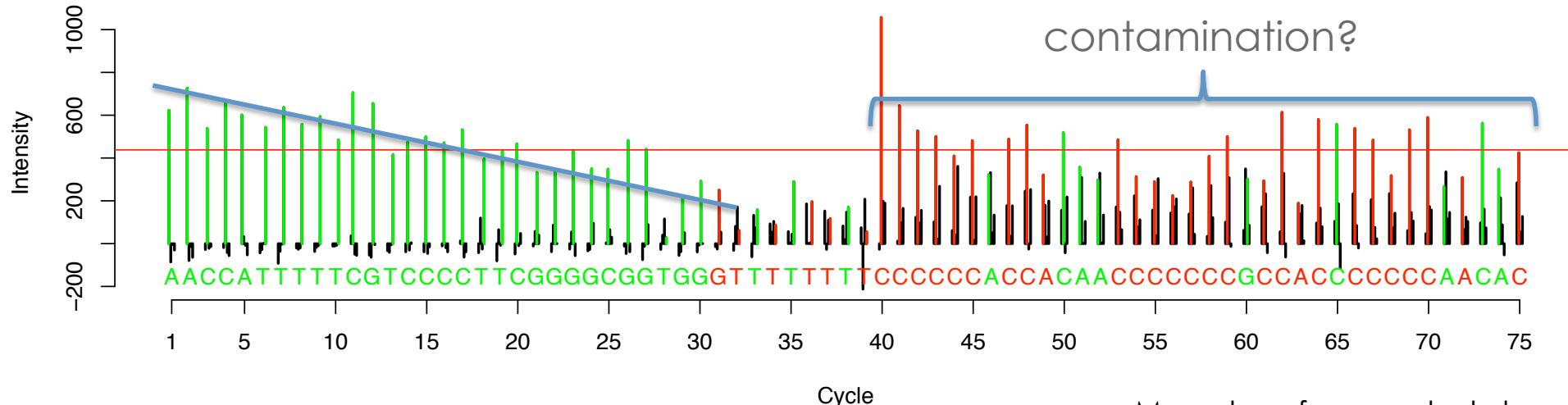
- ~ 45% of final cycle errors are due adapter sequence
- median purity still high; missing other effects?



AGATAGGAAGAGCGGTCACGCCACGAATGCCGAGA

Other attractions in the sequence zoo

Sick sequence: rapidly dies



Lazarus sequence: dies and rises again



AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Error frequencies

Manual look at all errors in 27 tiles of high quality sequence (Q34 bases)
Rates in Qphred

	Good read	Bad read	Indel	Adapter	Ligation	Unknown
Rate	36.9	40.1	47.7	46.3	47.0	45.6
Lower	36.5	39.5	46.3	45.1	45.7	44.5
Upper	37.4	40.8	49.4	47.8	48.5	46.9

Good read: only error is high quality base.

Bad read: otherwise messy, several errors.

Indel: presence of insertion or deletion

Adapter: undetected adapter sequence (after filtering)

Ligation: strong evidence of ligation

AGATAGGAAGAGCGGTTCAGGAAATGCCGAGA
Error frequencies

Manual look at all errors in 27 tiles of high quality sequence (Q34 bases)
Rates in Qphred

	Good read	Bad read	Indel	Adapter	Ligation	Unknown
Rate	36.9	40.1	47.7	46.3	47.0	45.6
Lower	36.5	39.5	46.3	45.1	45.7	44.5
Upper	37.4	40.8	49.4	47.8	48.5	46.9

$$\begin{array}{c}
 Q_{34} \\
 \downarrow \\
 \min(Q_{\text{Base}}, Q_{\text{Effect}}) \geq Q_{\text{Actual}} \geq \min(Q_{\text{Base}}, Q_{\text{Effect}}) - 3 \\
 \underbrace{\qquad\qquad}_{Q_{37} \geq} \qquad \qquad \qquad \underbrace{\qquad\qquad}_{Q_{37-3} \geq}
 \end{array}$$

AGATAGGAAGAGCGGTTCA~~GG~~CCACGGAAATGCCGAGA

Implementation and availability

<http://www.ebi.ac.uk/goldman/AYB/>

Written in R

Licensed under GPL (version 3)

Plug-in replacement for Bustard

- Single change to Makefile

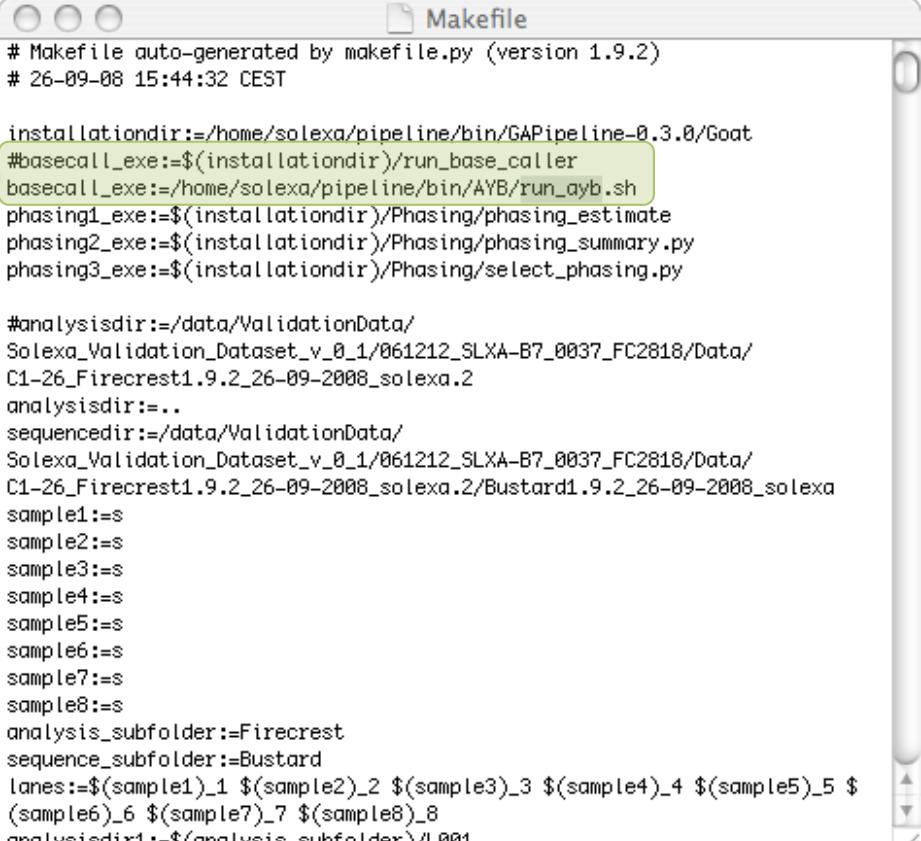
Change this line

Acceptable performance

2 hours per lane on an 8 core machine (~128 CPU hours per run)

- Faster if phasing and cross-talk assumed to constant (ala Bustard)
- Due to be rewritten with focus on performance and reliability

GAPipeline (Illumina) v. 0.3 Makefile



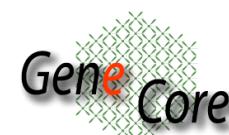
```
# Makefile auto-generated by makefile.py (version 1.9.2)
# 26-09-08 15:44:32 CEST

installationdir:=~/home/solexa/pipeline/bin/GAPipeline-0.3.0/Goat
#basecall_exe:=$(installationdir)/run_base_caller
basecall_exe:=~/home/solexa/pipeline/bin/AYB/run_ayb.sh
phasing1_exe:=$(installationdir)/Phasing/phasing_estimate
phasing2_exe:=$(installationdir)/Phasing/phasing_summary.py
phasing3_exe:=$(installationdir)/Phasing/select_phasing.py

#analysisdir:=~/data/ValidationData/
Solexa_Validation_Dataset_v_0_1/061212_SLXA-B7_0037_FC2818/Data/
C1-26_Firecrest1.9.2_26-09-2008_solexa.2
analysisdir:=..
sequencedir:=~/data/ValidationData/
Solexa_Validation_Dataset_v_0_1/061212_SLXA-B7_0037_FC2818/Data/
C1-26_Firecrest1.9.2_26-09-2008_solexa.2/Bustard1.9.2_26-09-2008_solexa
sample1:=s
sample2:=s
sample3:=s
sample4:=s
sample5:=s
sample6:=s
sample7:=s
sample8:=s
analysis_subfolder:=Firecrest
sequence_subfolder:=Bustard
lanes:=$(sample1)_1 $(sample2)_2 $(sample3)_3 $(sample4)_4 $(sample5)_5 \
$(sample6)_6 $(sample7)_7 $(sample8)_8
analysisdir1:=$(analysis_subfolder)/L001
```

Thanks to:

Jonathon Blake, EMBL Genomics Core Facilities

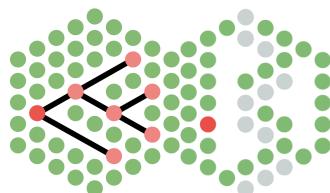


AGATAGGAAGAGCGGTTCA
CCACGGAAATGCCGAGA

Thanks

Thanks to:

- EBI:
 - Ewan Birney
 - Paul Flicek (1000 Genomes Project)
- EMBL Genomics Core Facilities, Heidelberg:
 - Vladimir Benes
 - Jonathon Blake
- Sanger Institute:
 - Nava Whiteford
(now at Oxford Nanopore Technologies)
 - Tom Skelly
 - Irini Abnizova
- Illumina:
 - Tony Cox
- CRUK Cambridge Research Institute:
 - Gordon Brown
 - Kevin Howe
- Cambridge Institute for Medical Research:
 - Vincent Plagnol
- Institute of Cell and Molecular Science,
Queen Mary University of London:
 - David Van Heel



君達の基地は、全て