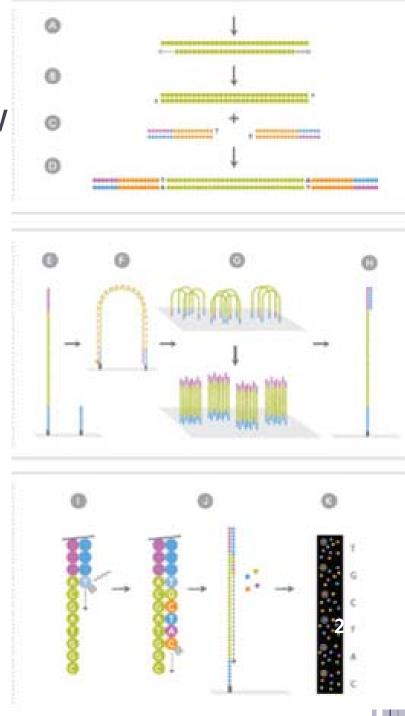
ALTA-CYCLIC: A SELF-OPTIMIZING BASE CALLER FOR NEXT-GENERATION SEQUENCING

Presenter: Jian Zhao

Illumina System Workflow

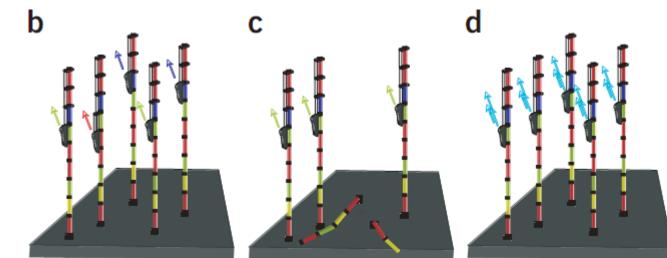
Library preparation
Cluster generation
Sequencing



Noise factors

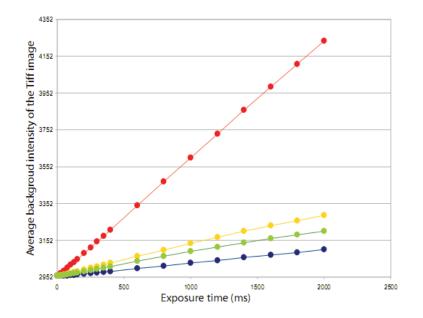
- Phasing noise
 - Leading, lagging
- Fading noise

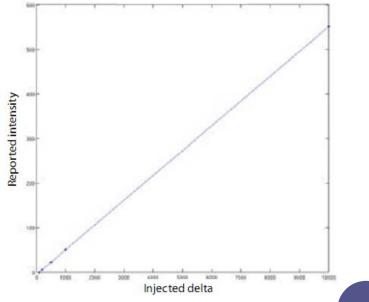
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- Exponential decay in fluorescent signal
- Cycle-dependent change in fluorophore crosstalk



Linearity of the intensity values

The optic chain is linear Firecrest applies linear transformation to the image





Impulse response analysis

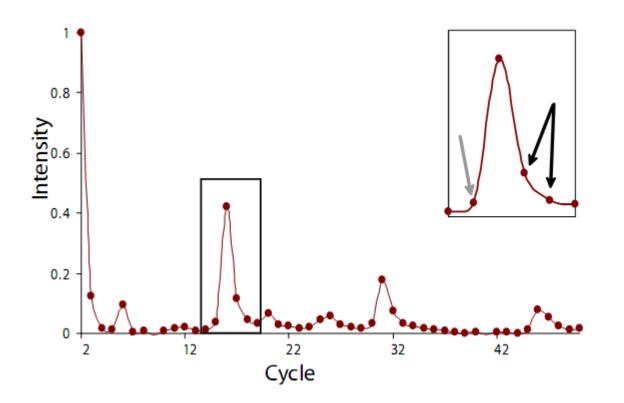
• Synthesized DNA fragments

- Delta function
- Dinucleotide microsatellites
- Theta function

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∆ on randm context	C <u>A</u> GTCGGCCGTCGGT <u>A</u> TCCTGGTGGTGGCT <u>A</u> GGCTGTCTCTTTCC <u>A</u> CGGC G <u>C</u> AGTAGTGTTGGTT <u>C</u> TGTAGTGGAATGTG <u>C</u> GGTTGTTGAGAATT <u>C</u> AGTA C <u>G</u> CCTTACAATTCAAAGTCCATATAACTTT <u>G</u> AATAACCTTACATC <u>G</u> ATAT C <u>T</u> AGCCGCGACAACA <u>T</u> AGCAGGCACGAGAG <u>T</u> CGACGGACAGCGGA <u>T</u> GCGA
∆ on homeo- polymer con- text	A <mark>C</mark> AAAAAAAAAAACCAAAAAAAAAAAAAAAAAAAAAA
MS	ACACACACACACACACACACACACACACACATTG GTGTGTGTGTGTGTGTGTGTGTGTGTGTGTGTGTG
θ	AAAAAAACCCCCCCAAAAAAACCCCCCCAAAAAAACCCC GGGGGGGG

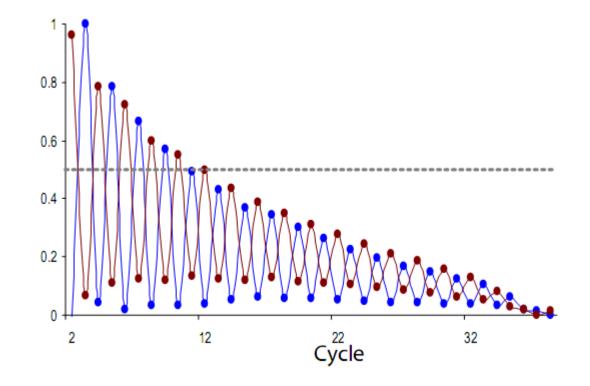
Noise factors - phasing

 Impulse response test of delta function: GCAGTAGTGTTGGTTCTGTAGTGGAATGTGCGGTT GTTGAGAATTCAGTA



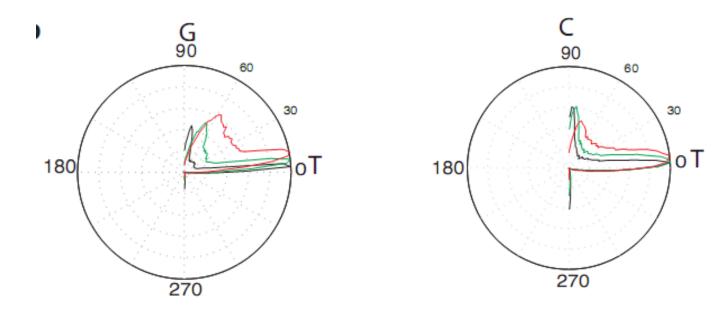
Noise factors - fading

Output average intensities of microsatellite sequence ACAC...



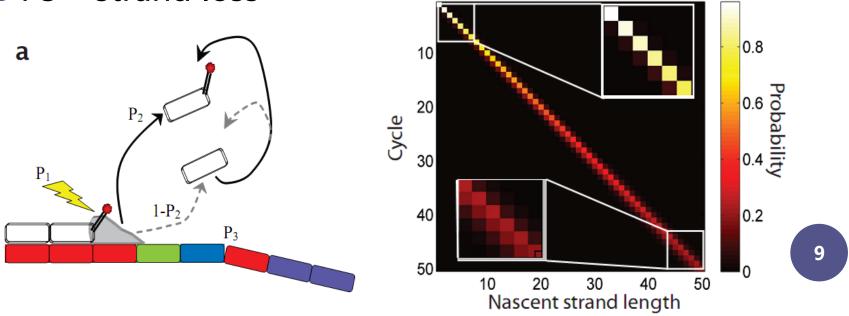
Noise factors – crosstalk change

 Polar histograms present the ratio between channel intensities correlated with the base preference (bacteriophage phi-X library)



Random walk model of phasing & fading

- P1 block removal
 - stay the same length with 1-P1
- P2 incorporation of blocked nucleotide
 - incorporation of non-blocked nucleotide with 1-P2
- P3 strand loss



Decomposition of phasing & fading

DP = R

- R(t, n) probability of a nascent strand to be n nucleotides long after t cycles
- D fading matrix (t by t diagonal)
- P phasing matrix (t by n)
 - P(t, n) probability of finding a nascent strand with length n after t cycles

Intensity of DNA cluster

$$\eta_j \cdot DPS_j G^T = I_j$$

η_j – size of j-th DNA cluster (scalar)
S_j – DNA sequence of j-th cluster (n by 4)
G – crosstalk matrix (4 by 4)
I_i – intensity signal of j-th cluster (t by 4)

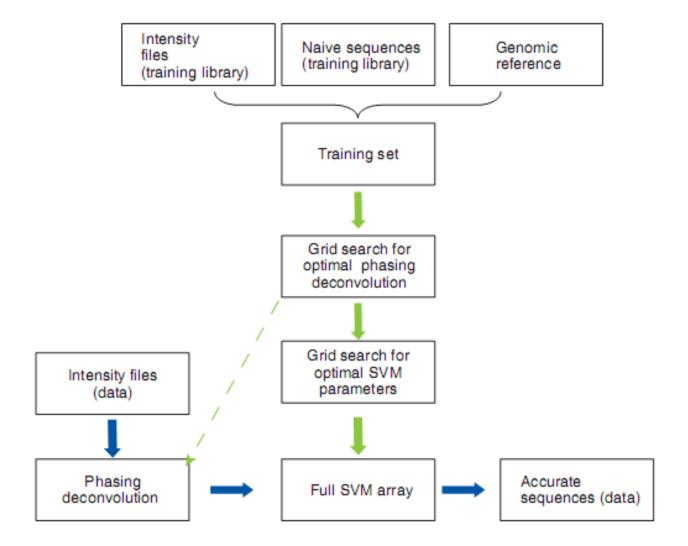
$$\eta_j \cdot (PD)^+ DPS_j G^T G^{-T} = (PD)^+ I_j G^{-T}$$
$$\eta_j \Sigma S_j = Y$$

Note the crosstalk matrix G is cycle-dependent!!!

Alta-Cyclic

- Treat sequencing as a classification problem, use SVM to learn noise patterns
- Training set: fluorescence intensities and corresponding correct base calls
- Training process
 - Deconvololute the phasing effect of intensities according to grid coordinate
 - Pick the intensities and correct base calls of last few cycles, run SVM for each cycle
 - Average success rate in the cross-validation of SVMs is used as a feedback to the grid search
 - Optimize SVM parameters by using grid search

Steps of training and base calling

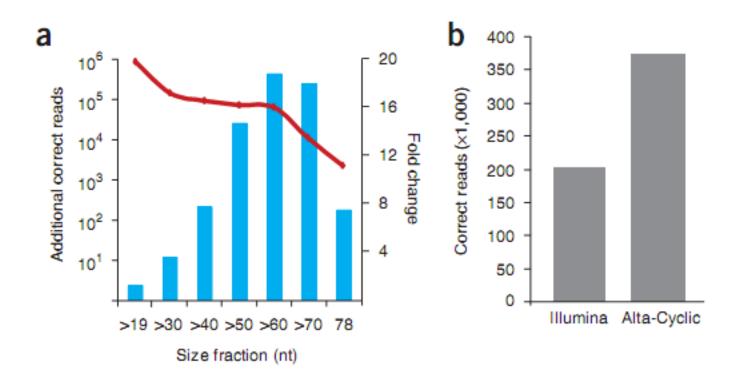


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Features of Alta-Cyclic
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- All the calling parameters are optimized empirically and tested to enhance the accuracy for each run
- Phasing parameters are based on a parametric model and calculated from data of latest cycles
- Dynamically track changes in fluorophore cross-talk

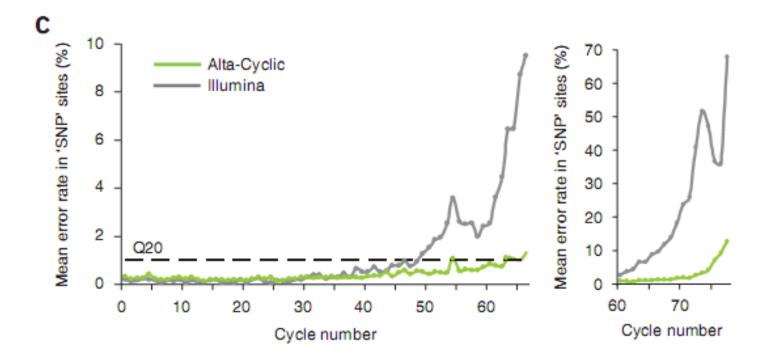
Experiments - for long runs

A: HepG2 RNA libraryB: Tetrahymena micronuclear library



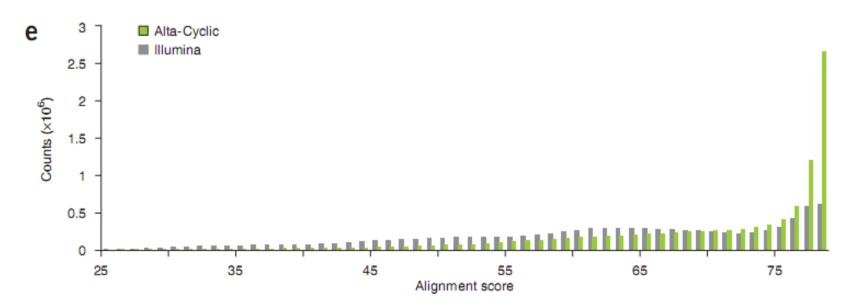
Experiments – identify sequence variants

 Phi X library with 1% artificially single-base changes



Experiments – very noisy reads

Align output to phi x genome (allowing 53 mismatches out of 78)



Weak points (from my point of view)

- Need more computation time; iterative grid search of parameters is time consuming
- Referencing DNA library must be prepared and extra DNAs must be sequenced for each run
- Training dataset could be noisy
- SVM parameters used in grid search for parameters of random walk model are not mentioned.

