

Read Orientation Artifact Filter for Somatic Variants

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DSDE Methods Meeting
9/22/17

What is read orientation

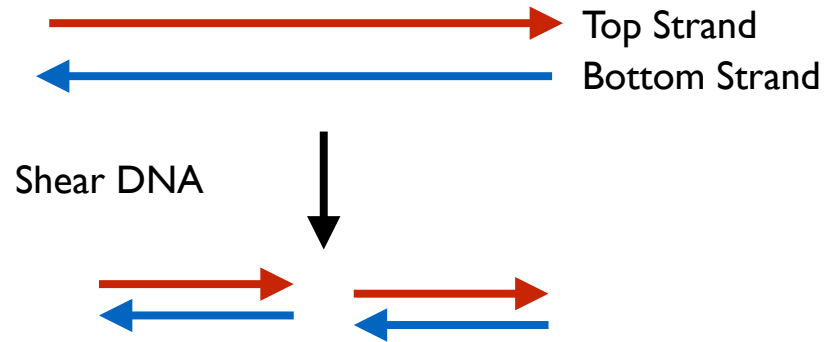
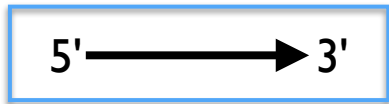
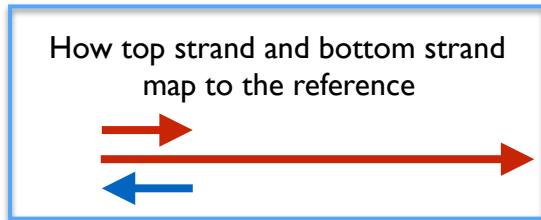
How top strand and bottom strand
map to the reference



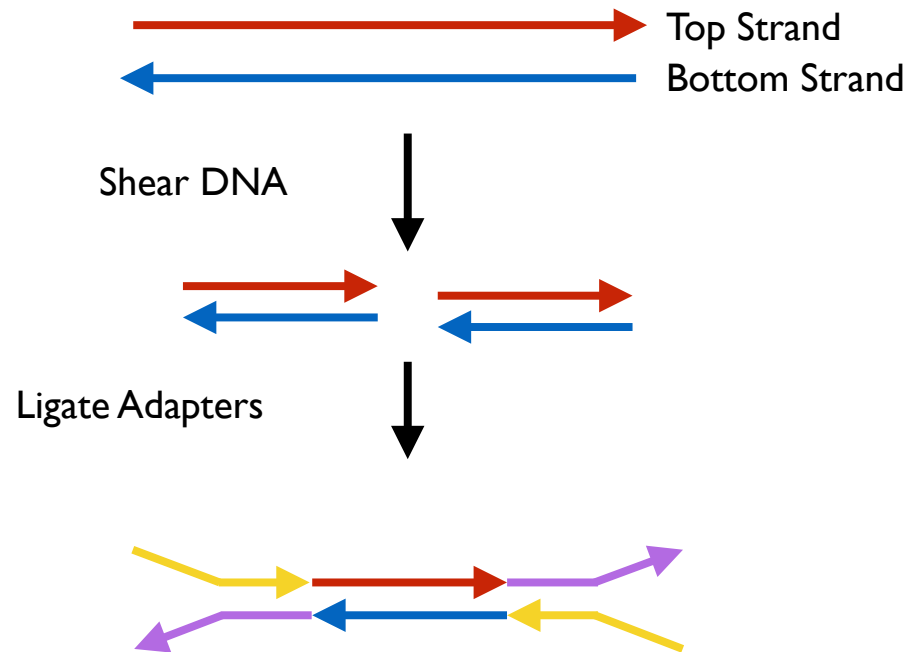
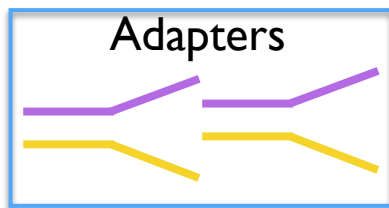
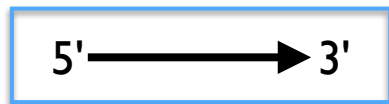
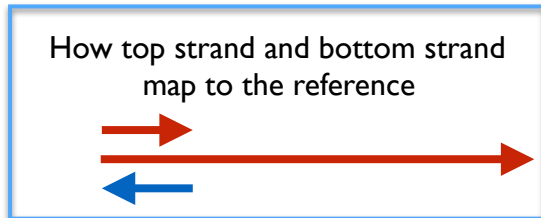
5' → 3'



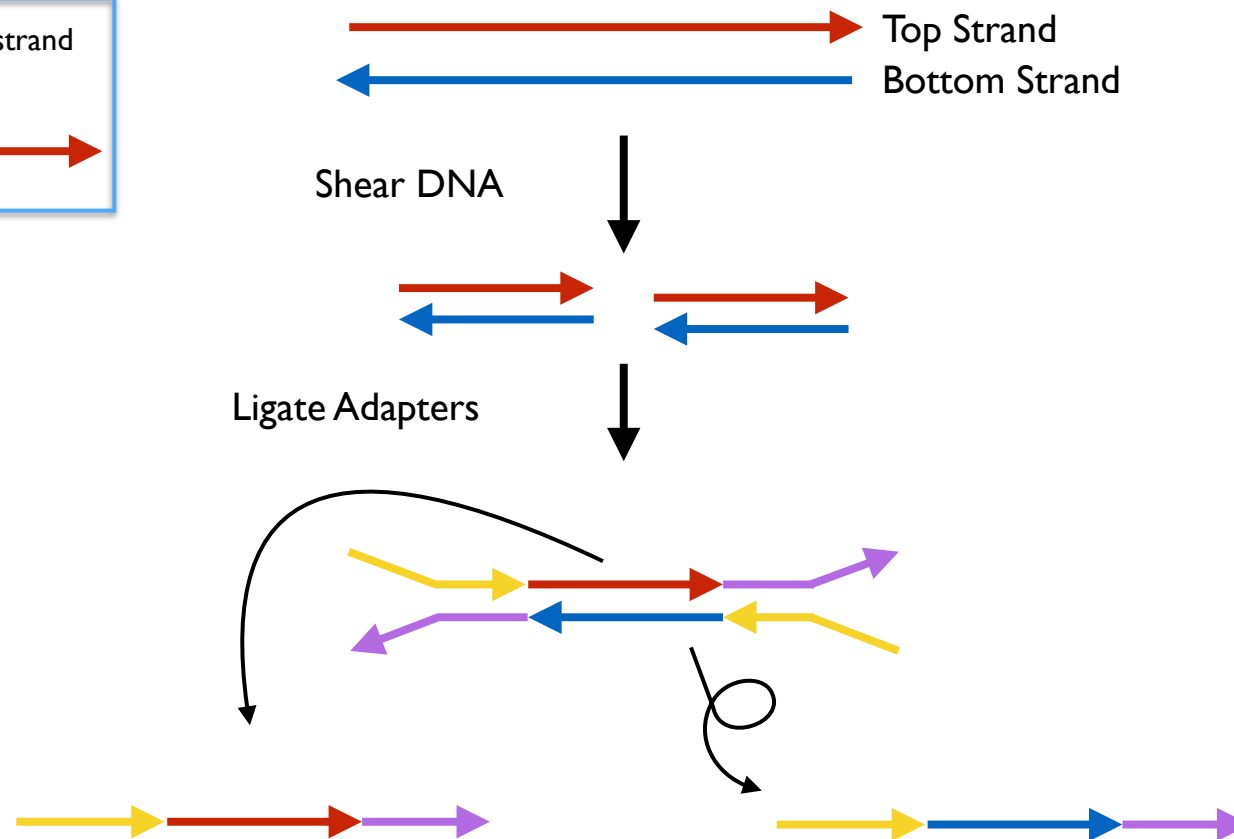
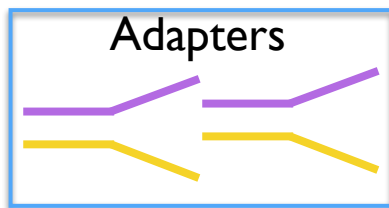
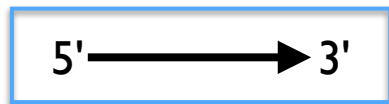
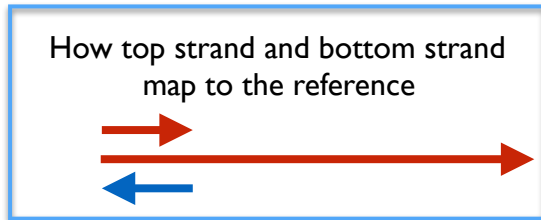
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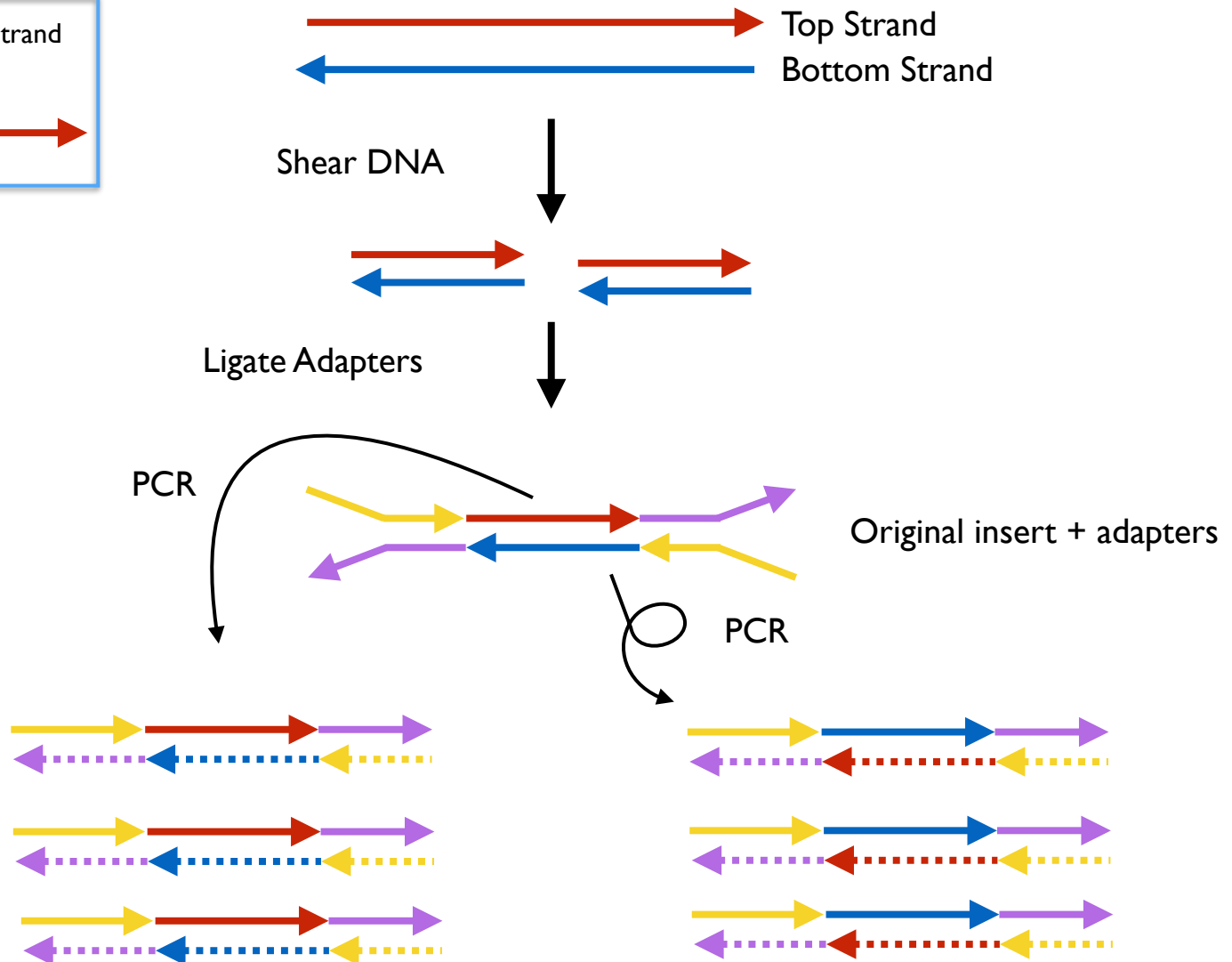
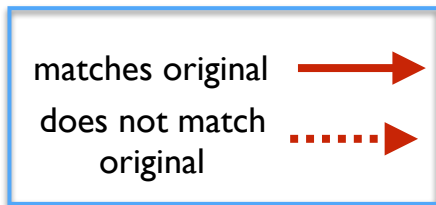
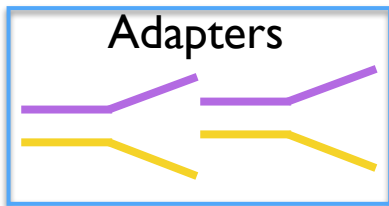
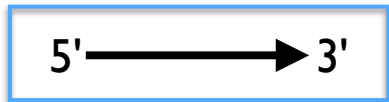
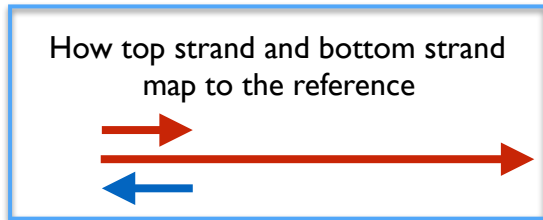
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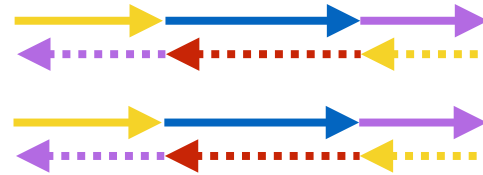
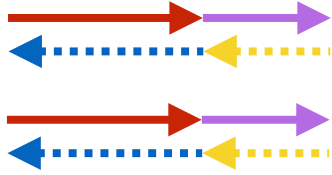
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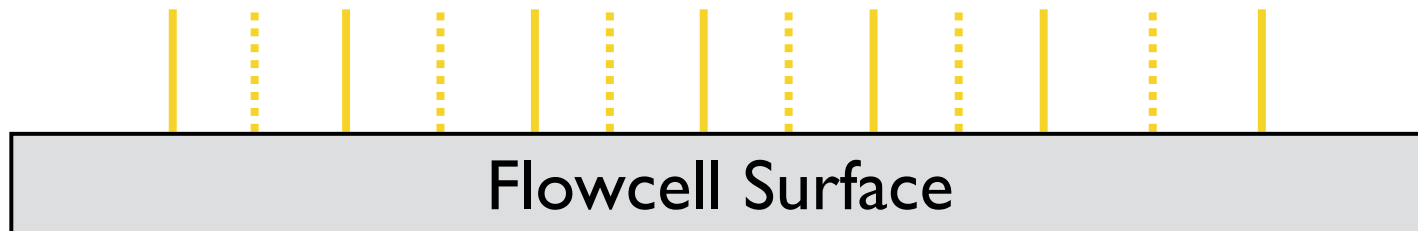
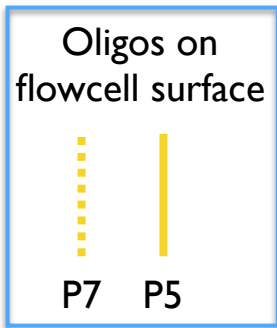
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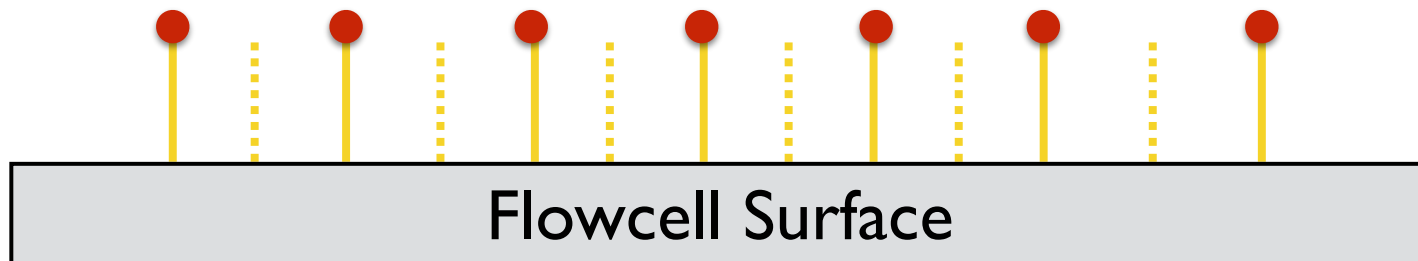
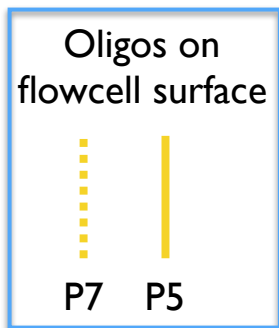
Illumina Sequencing



Illumina Sequencing

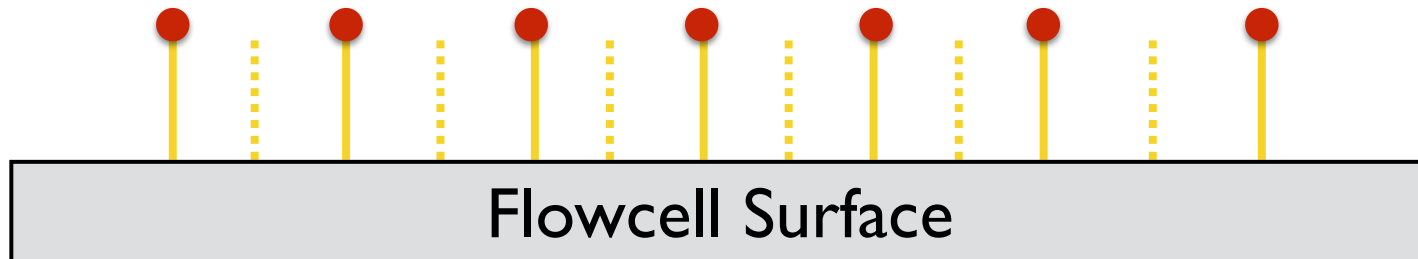
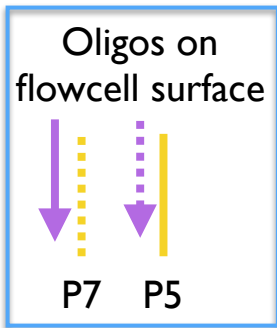


Illumina Sequencing

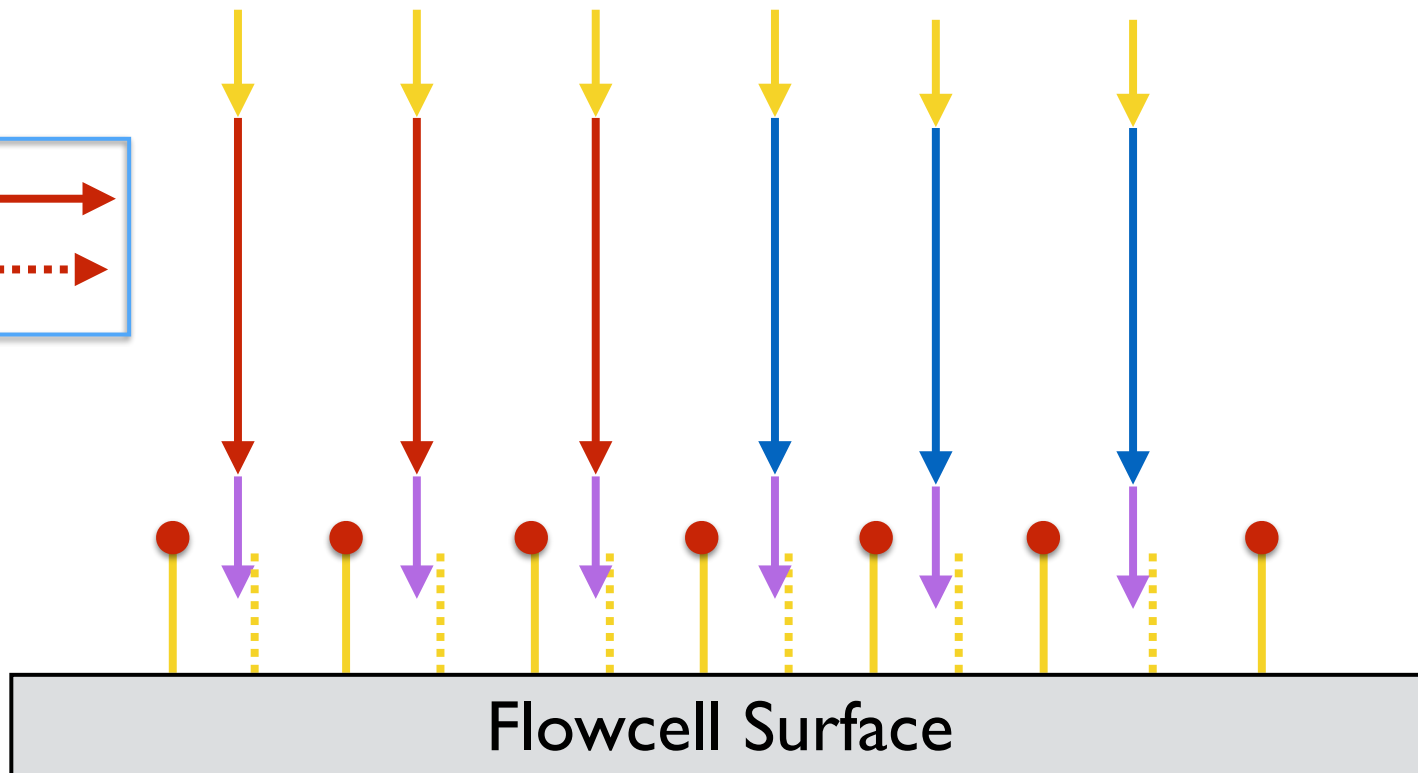
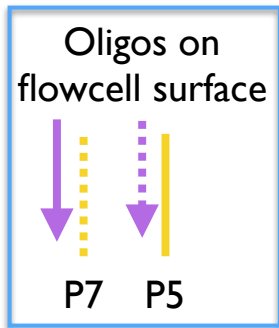
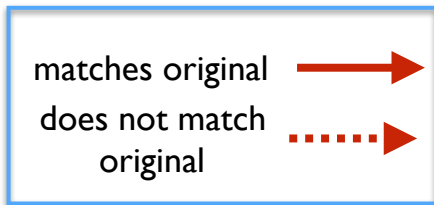


Block the p5 oligos

Illumina Sequencing

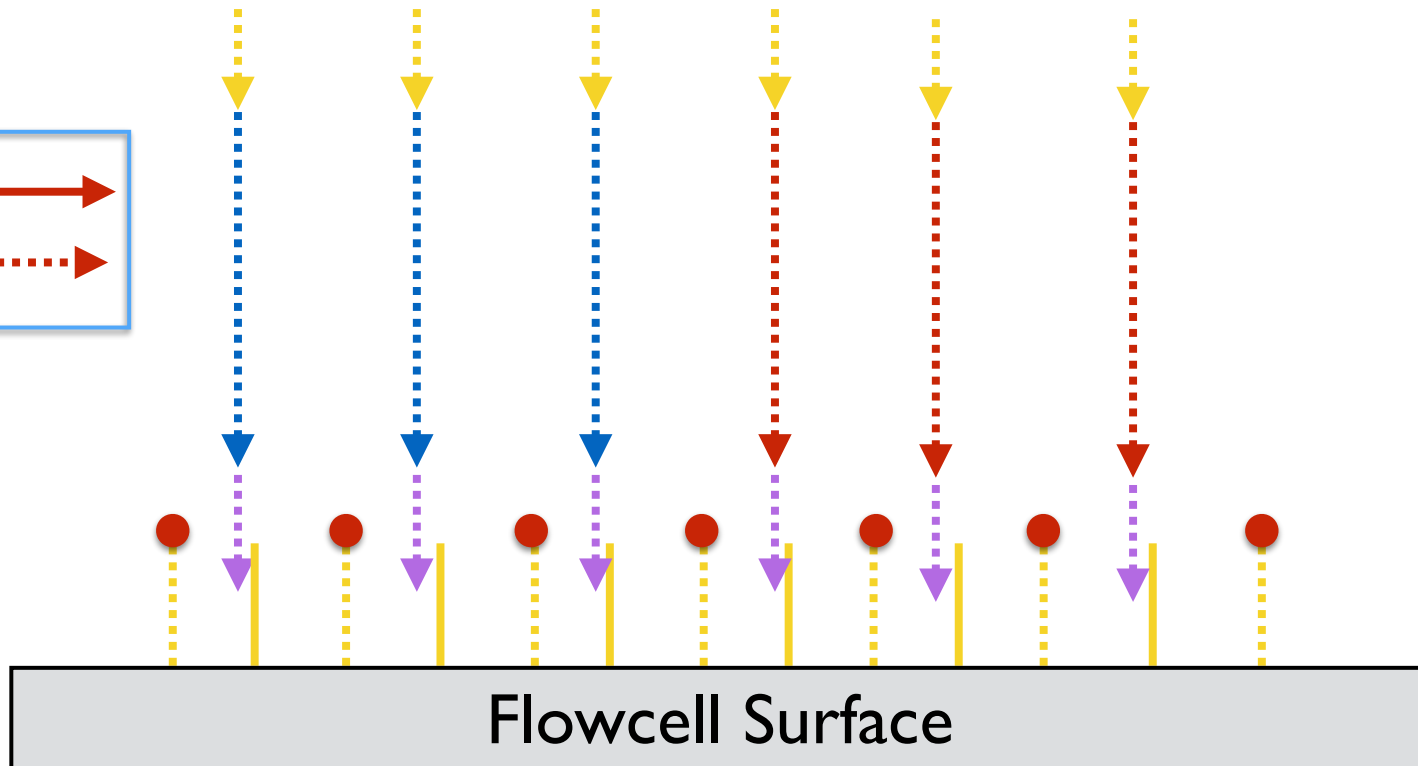
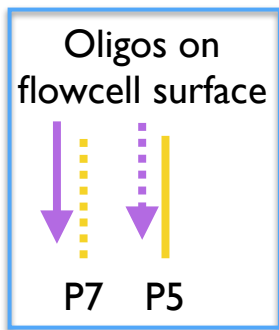
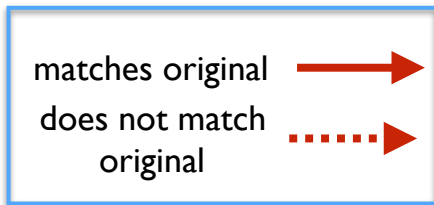


Illumina Sequencing



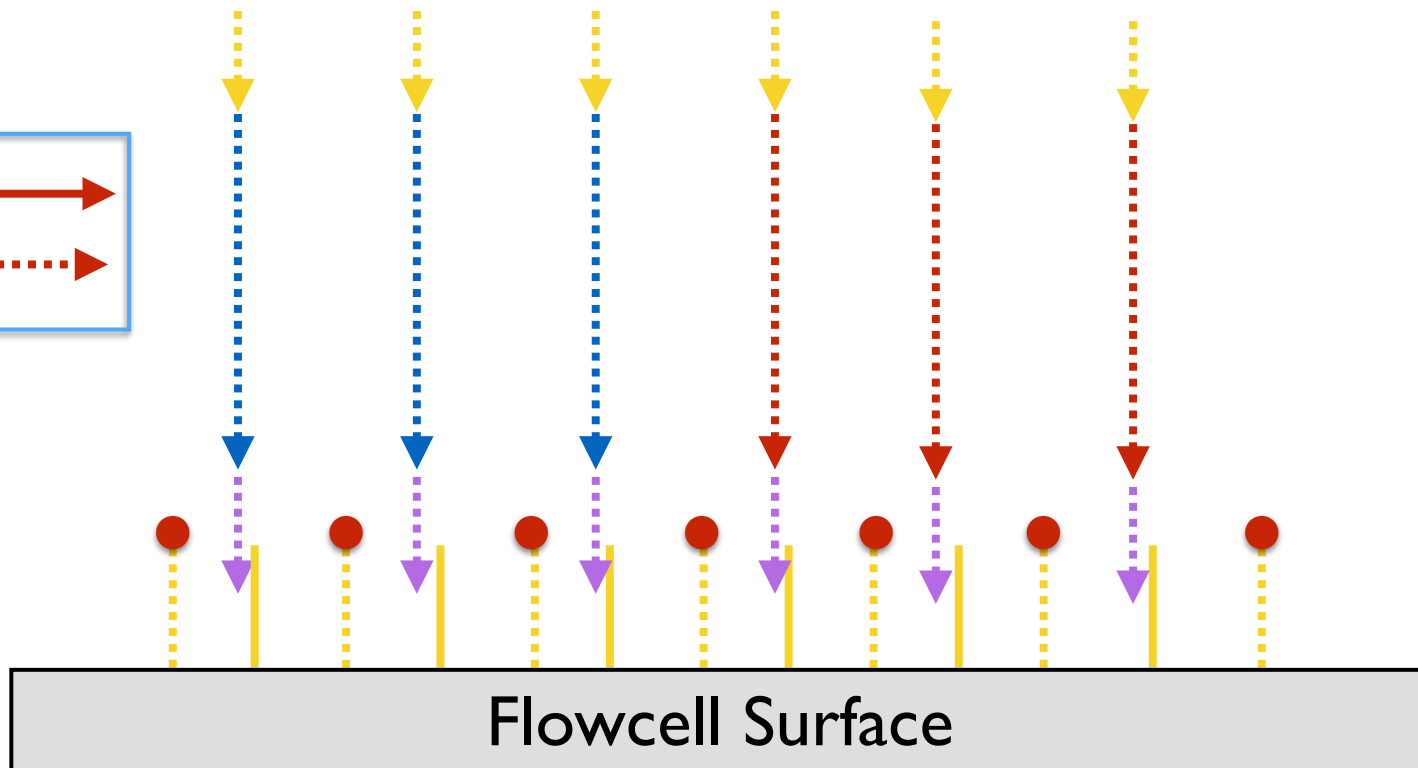
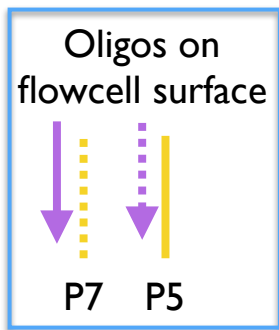
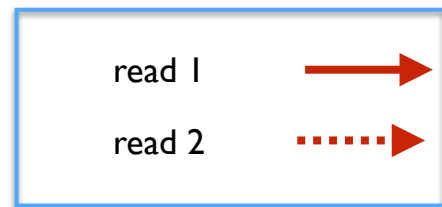
↓ Read 1: use solid adapter

Illumina Sequencing



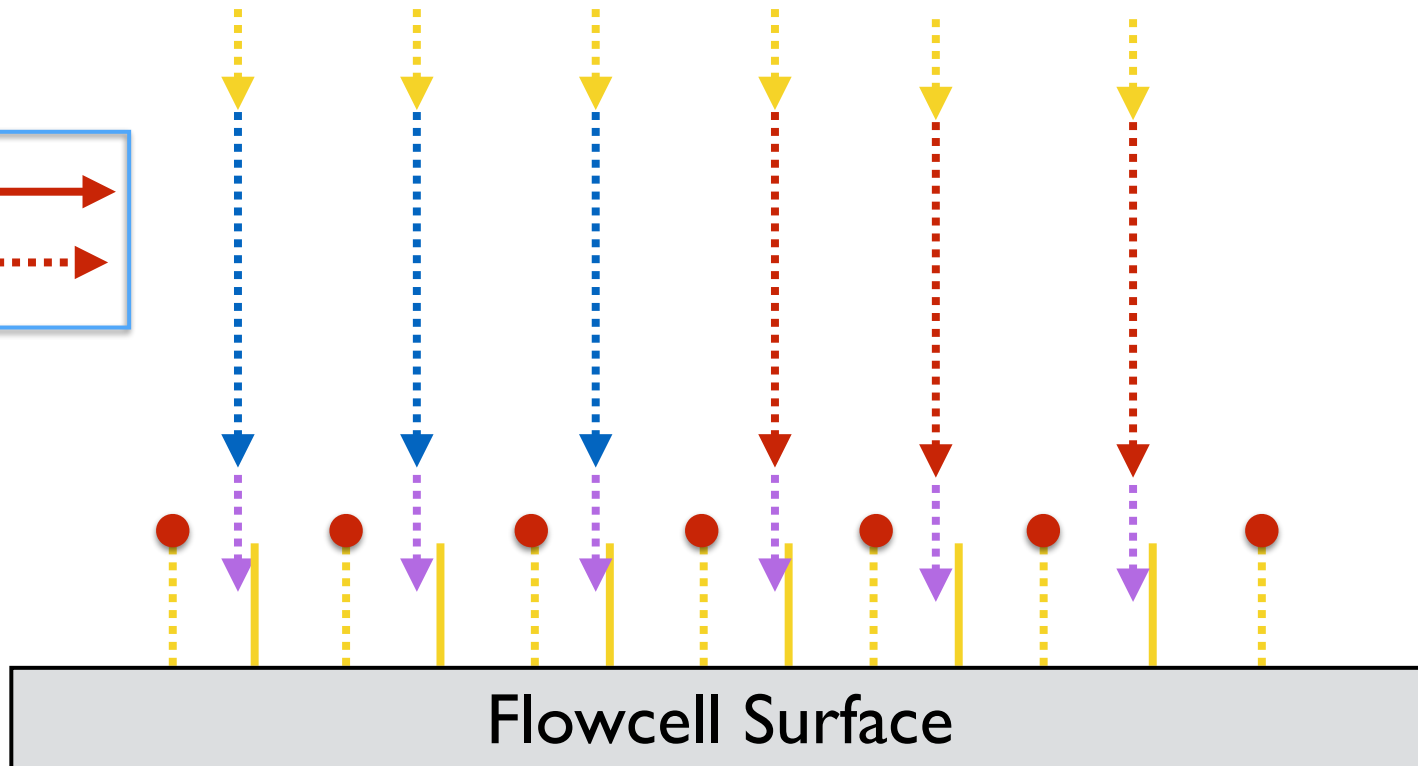
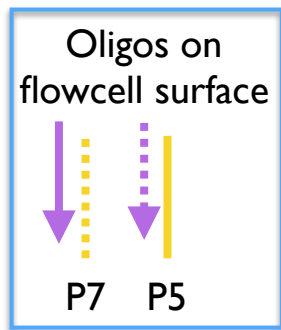
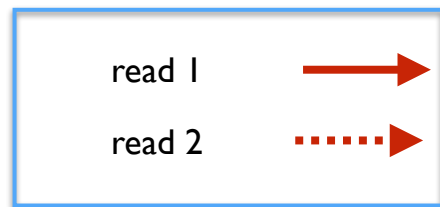
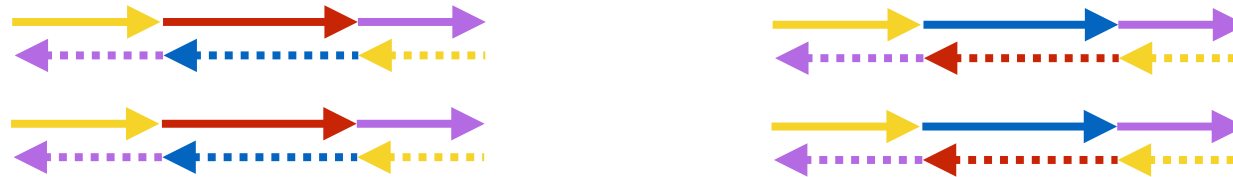
Read 2: use dotted adapters (not original adapter seq)

Illumina Sequencing



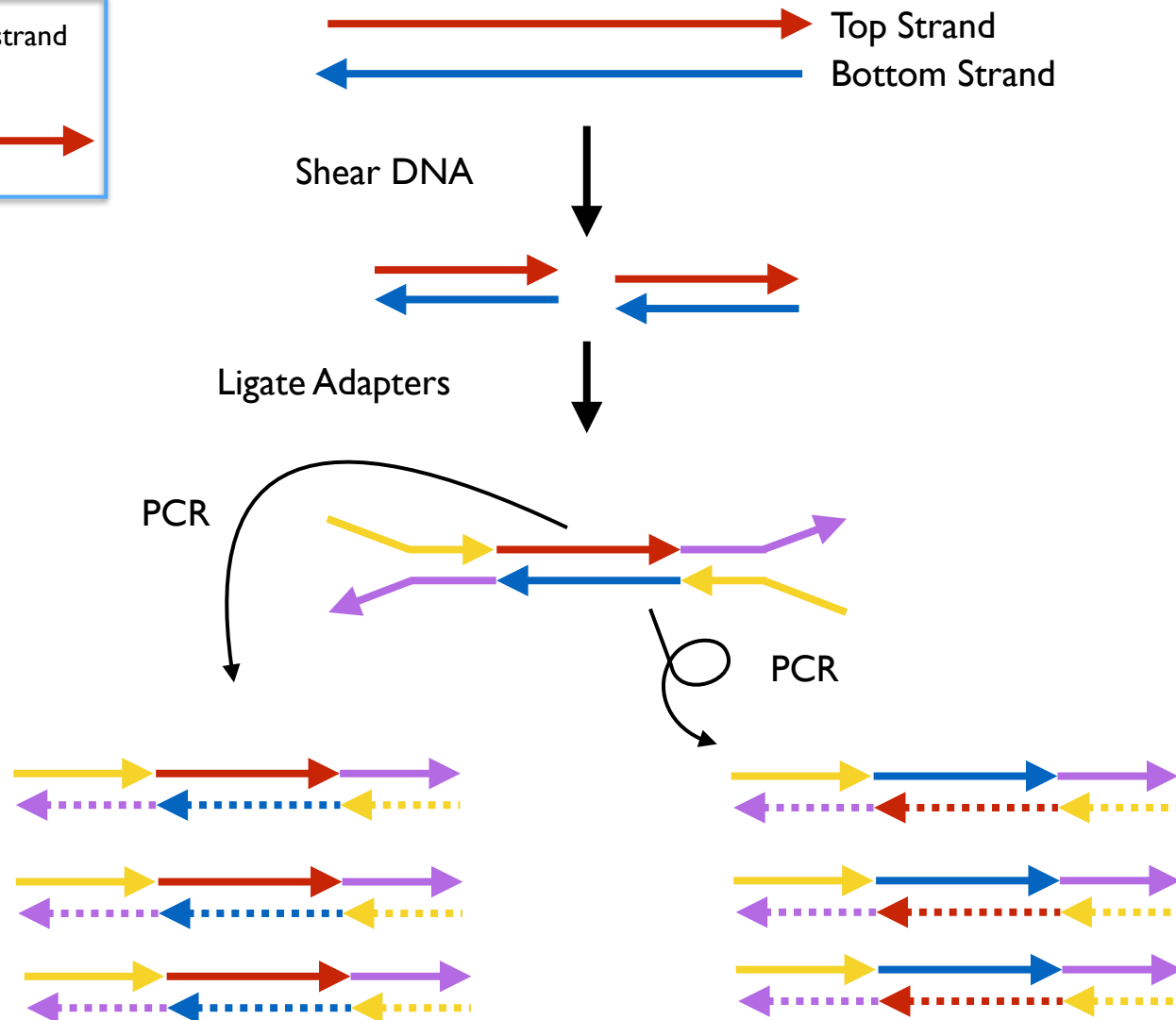
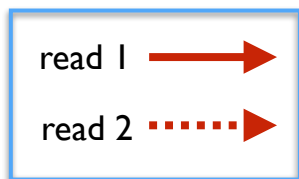
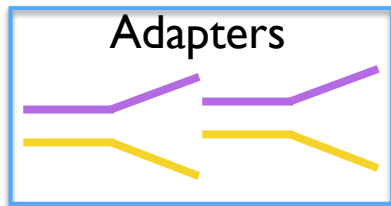
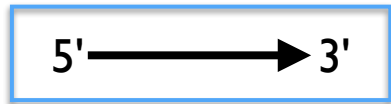
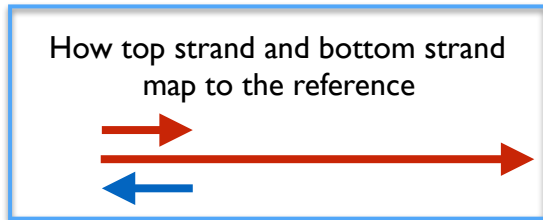
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Illumina Sequencing

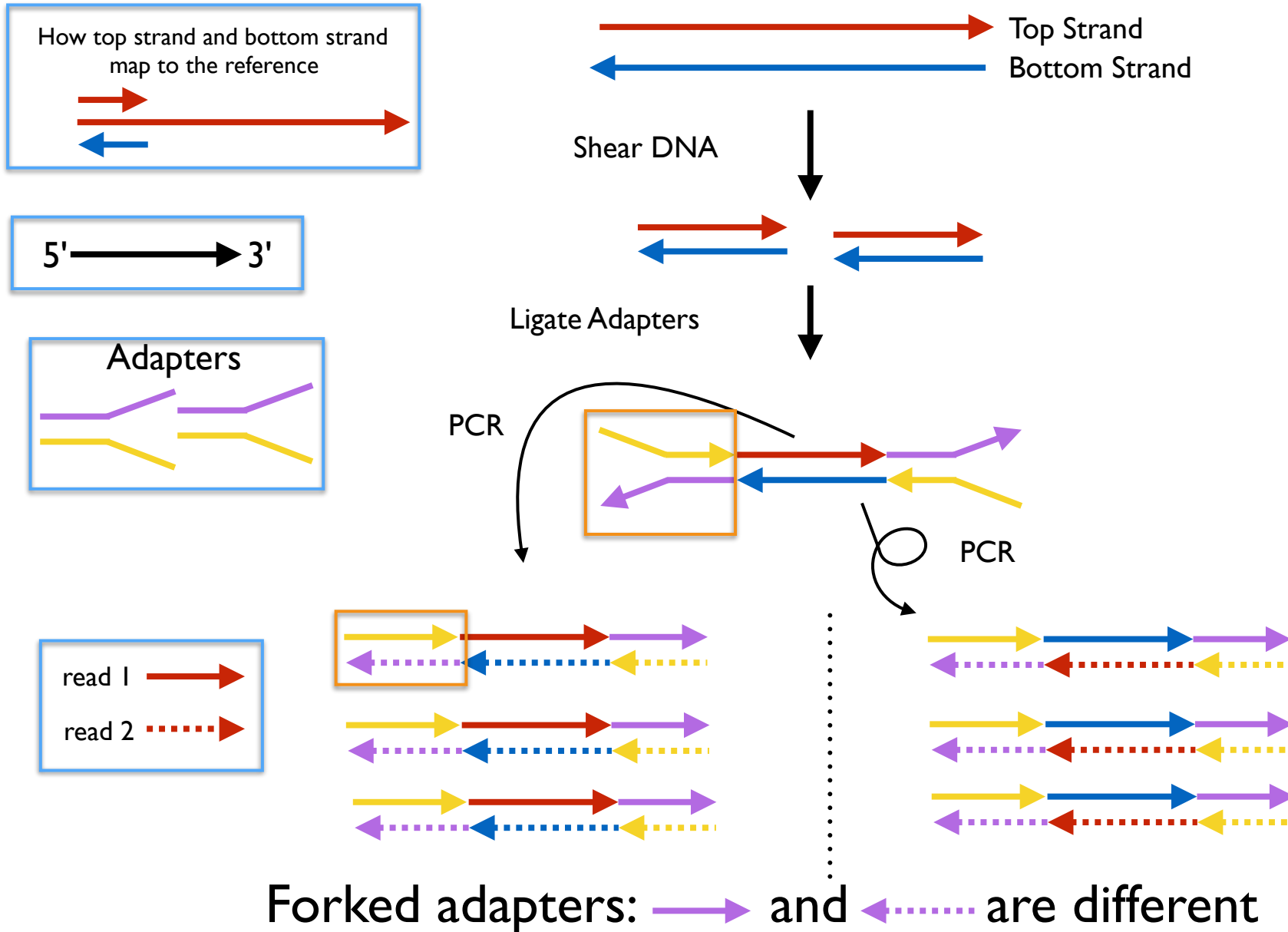


Aren't  and  identical?

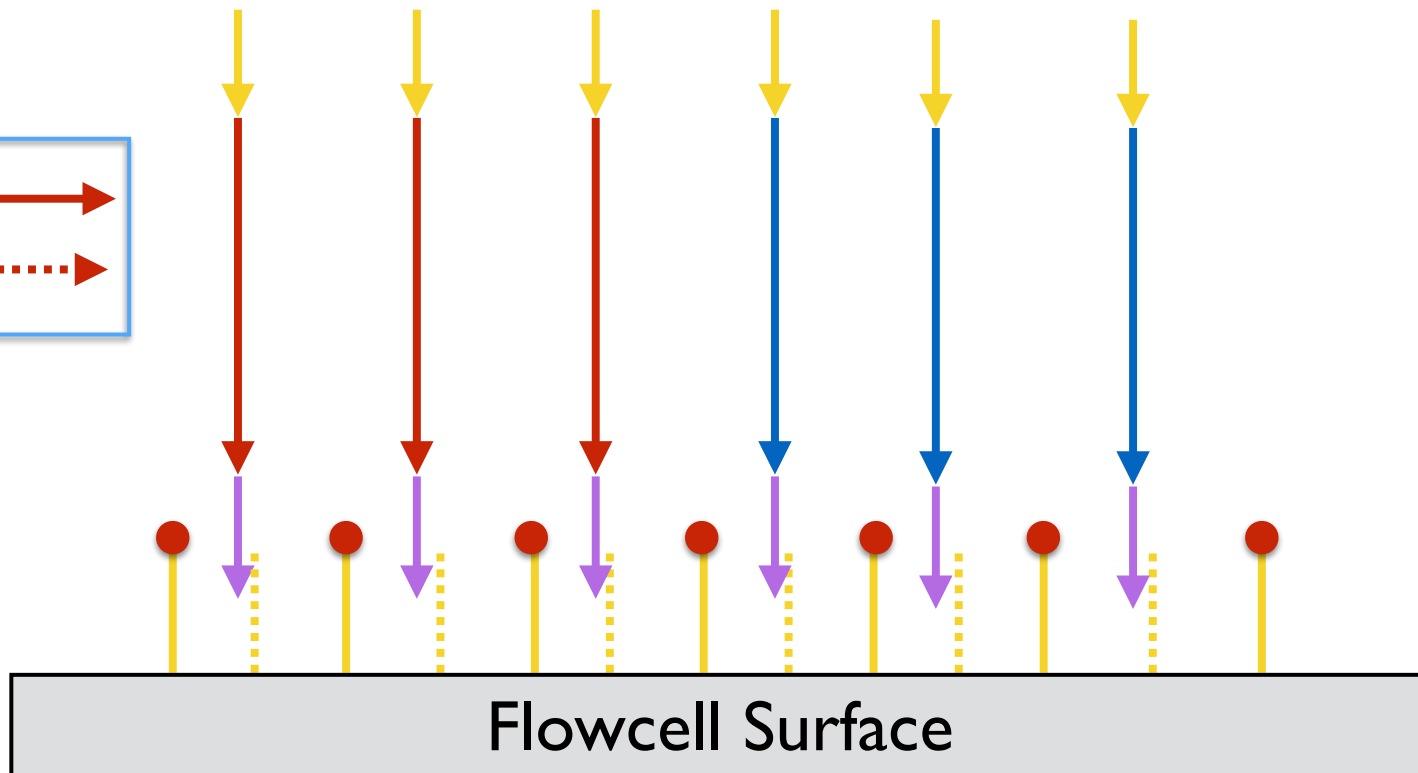
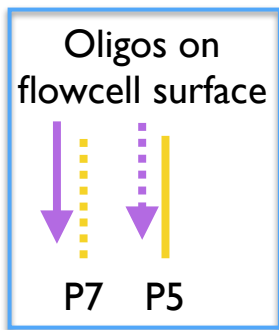
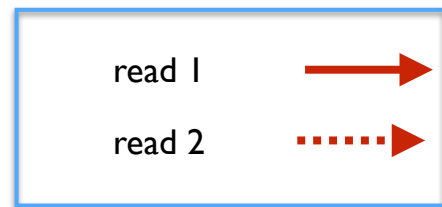
Recall...



Recall...forked adapters



Illumina Sequencing



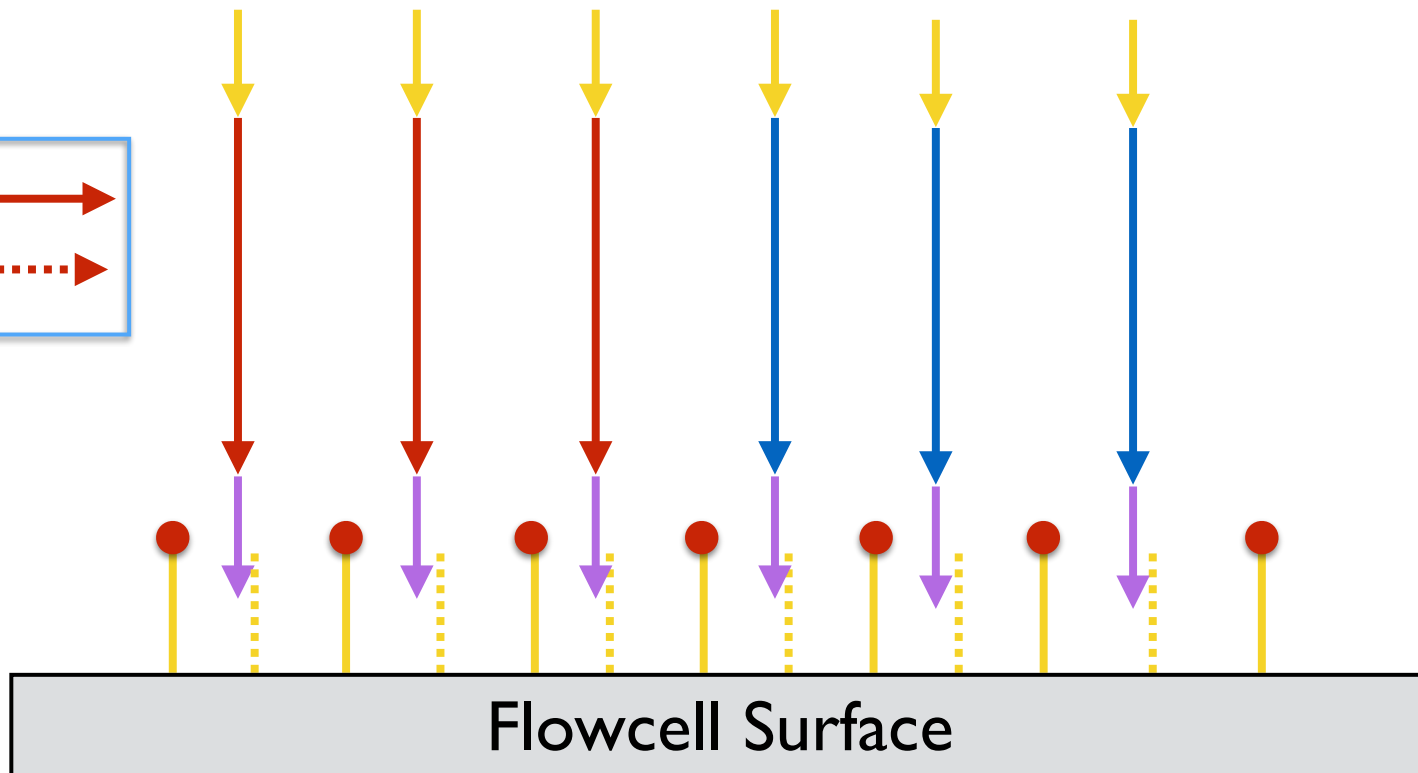
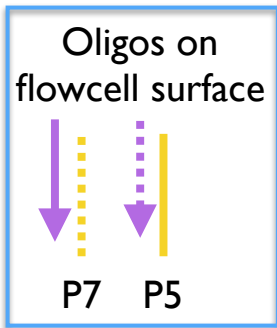
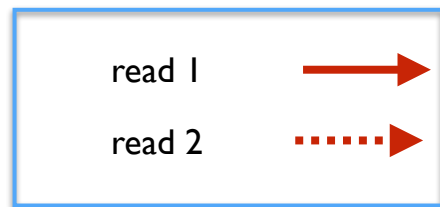
Illumina Sequencing



Read 1 = top strand



Read 1 = bottom strand

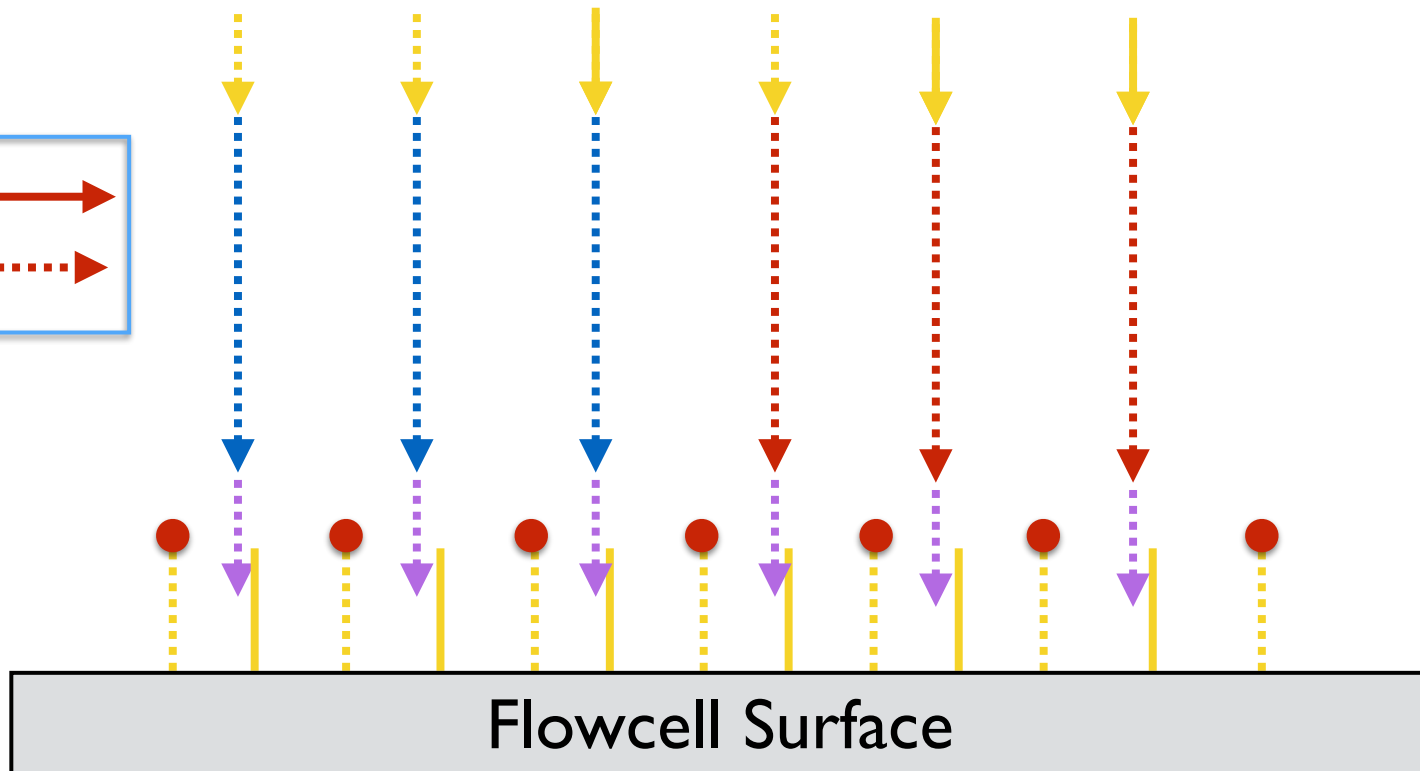
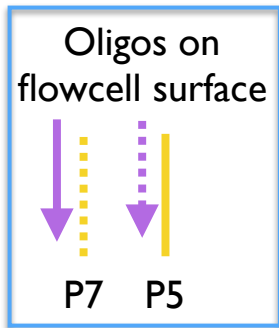
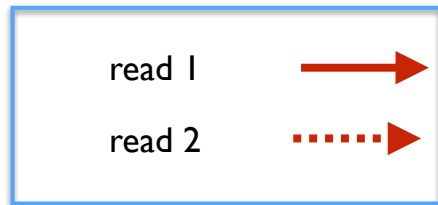


Illumina Sequencing



Read 1 = top strand
Read 2 = bottom str.

Read 1 = bottom strand
Read 2 = top strand

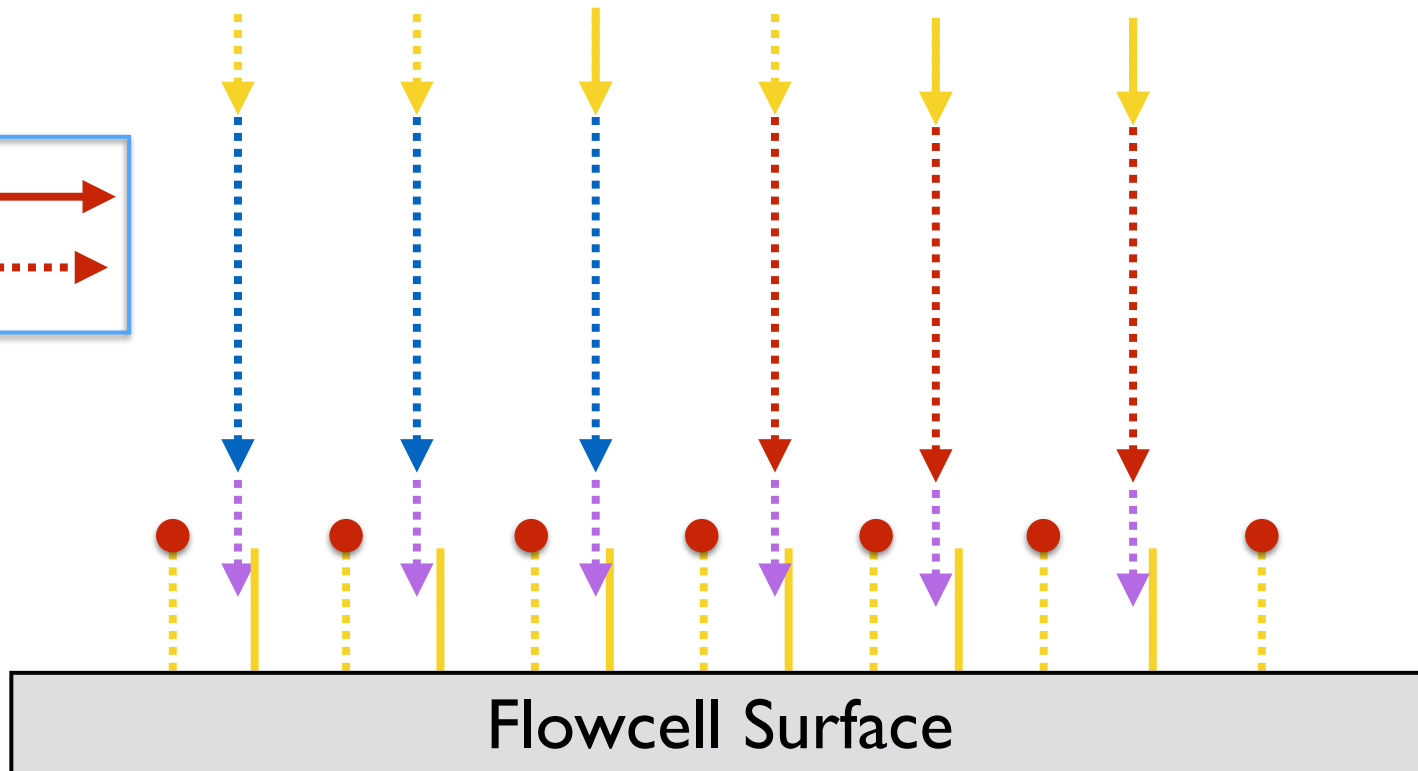
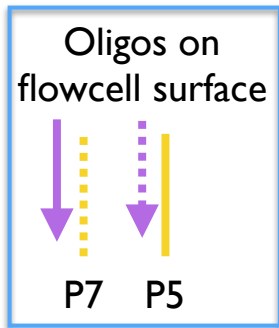
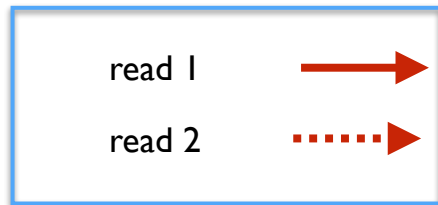


Illumina Sequencing

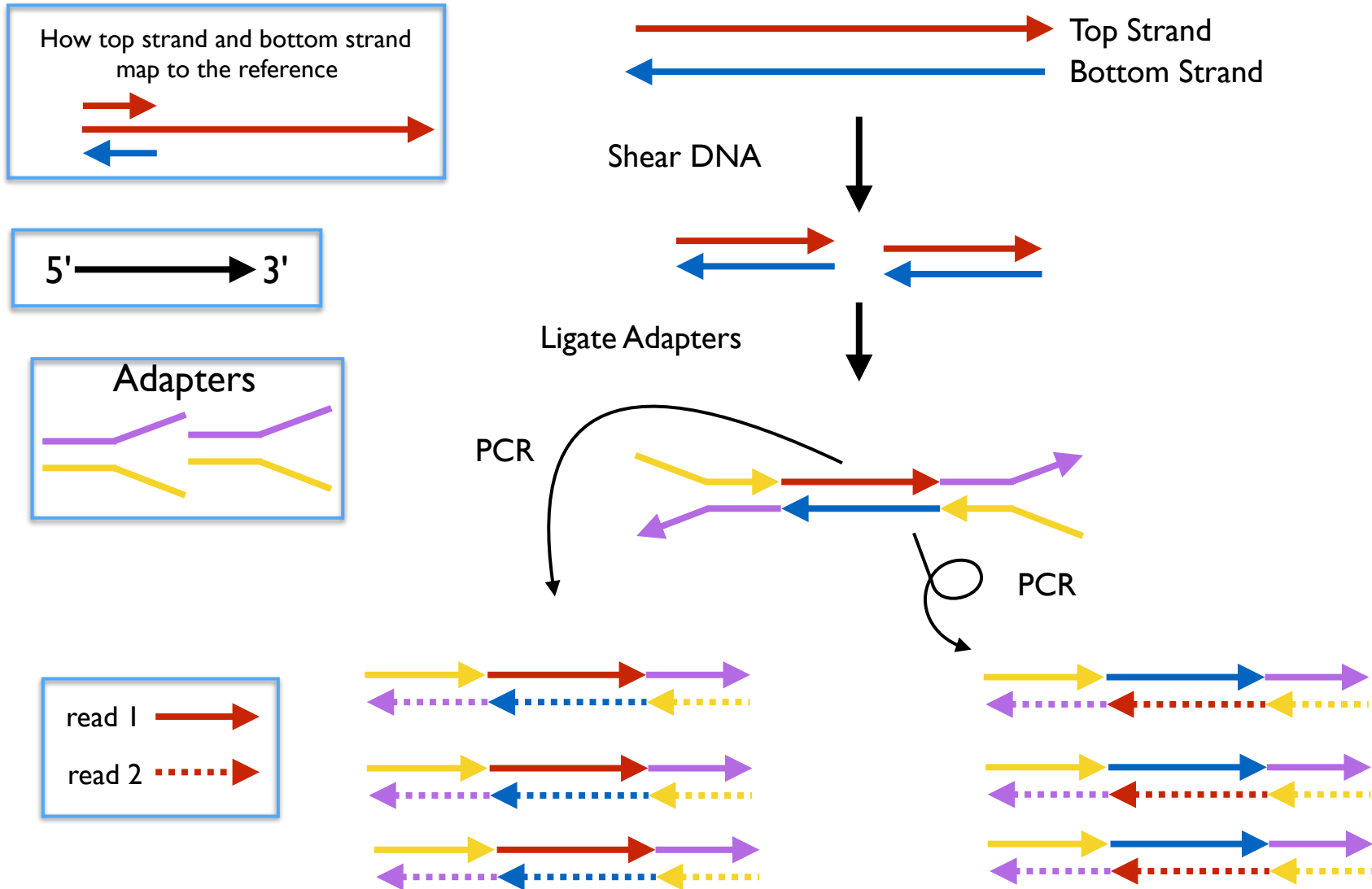


FIR2

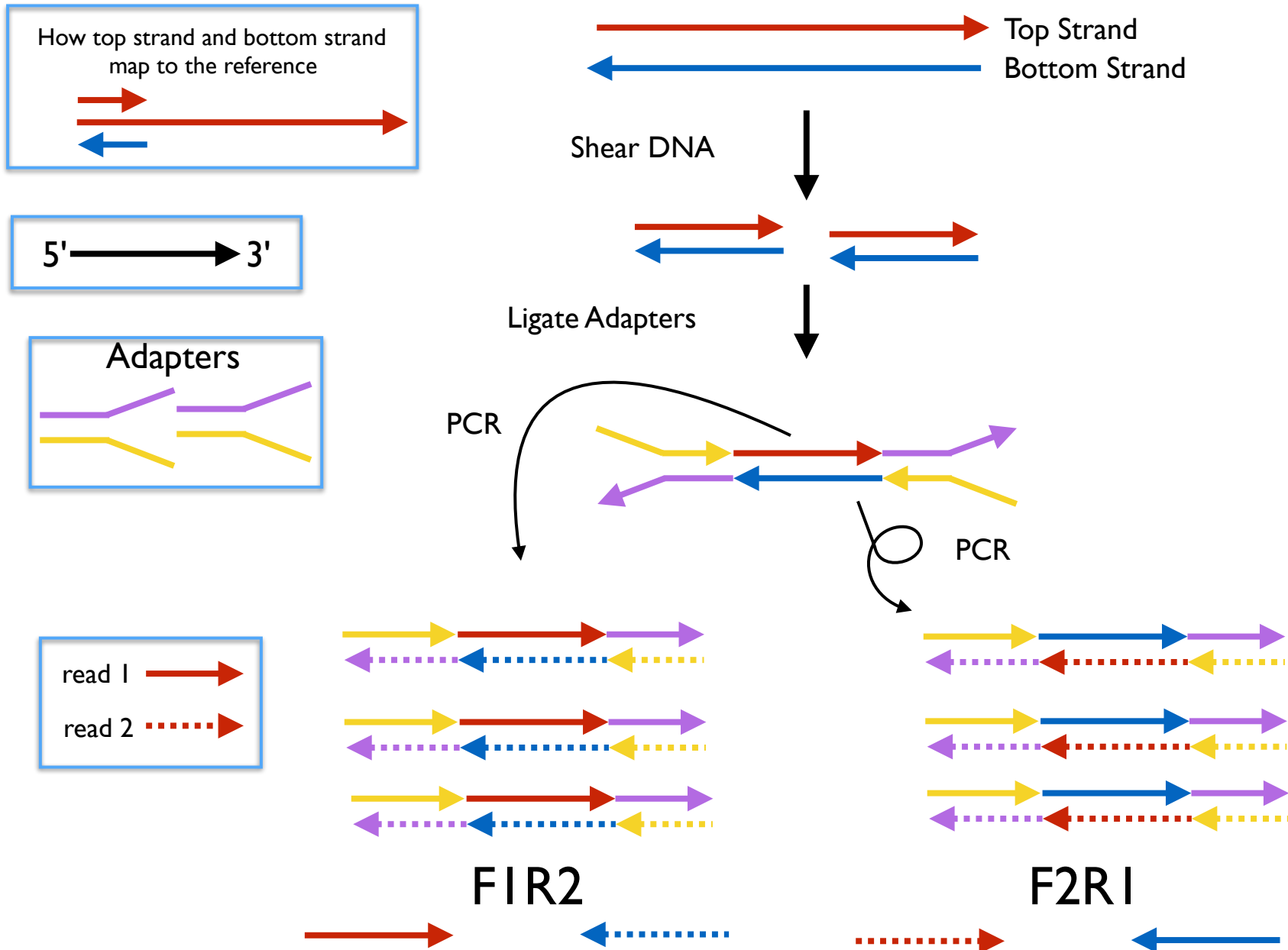
F2R1



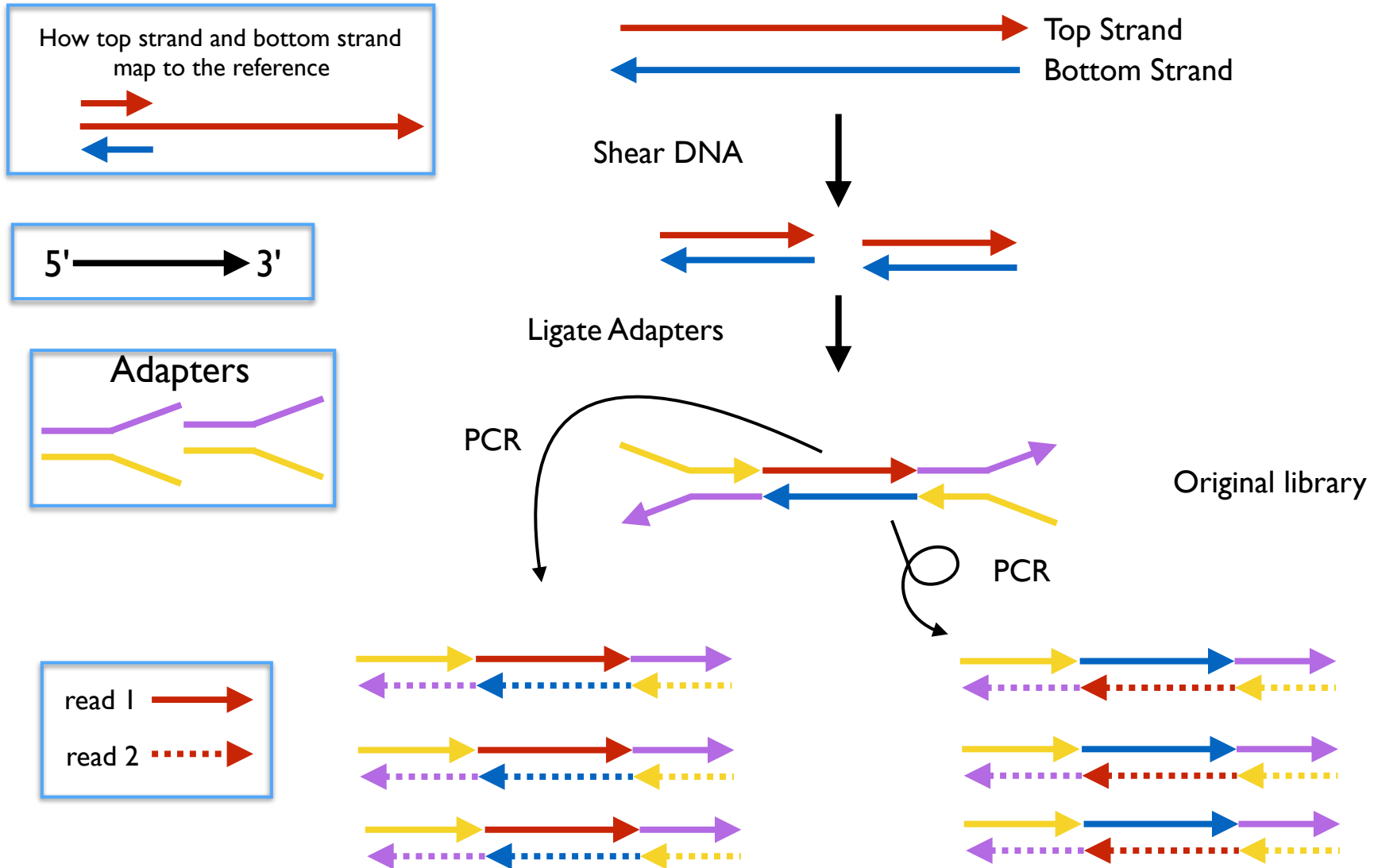
What is read orientation



What is read orientation



What is read orientation

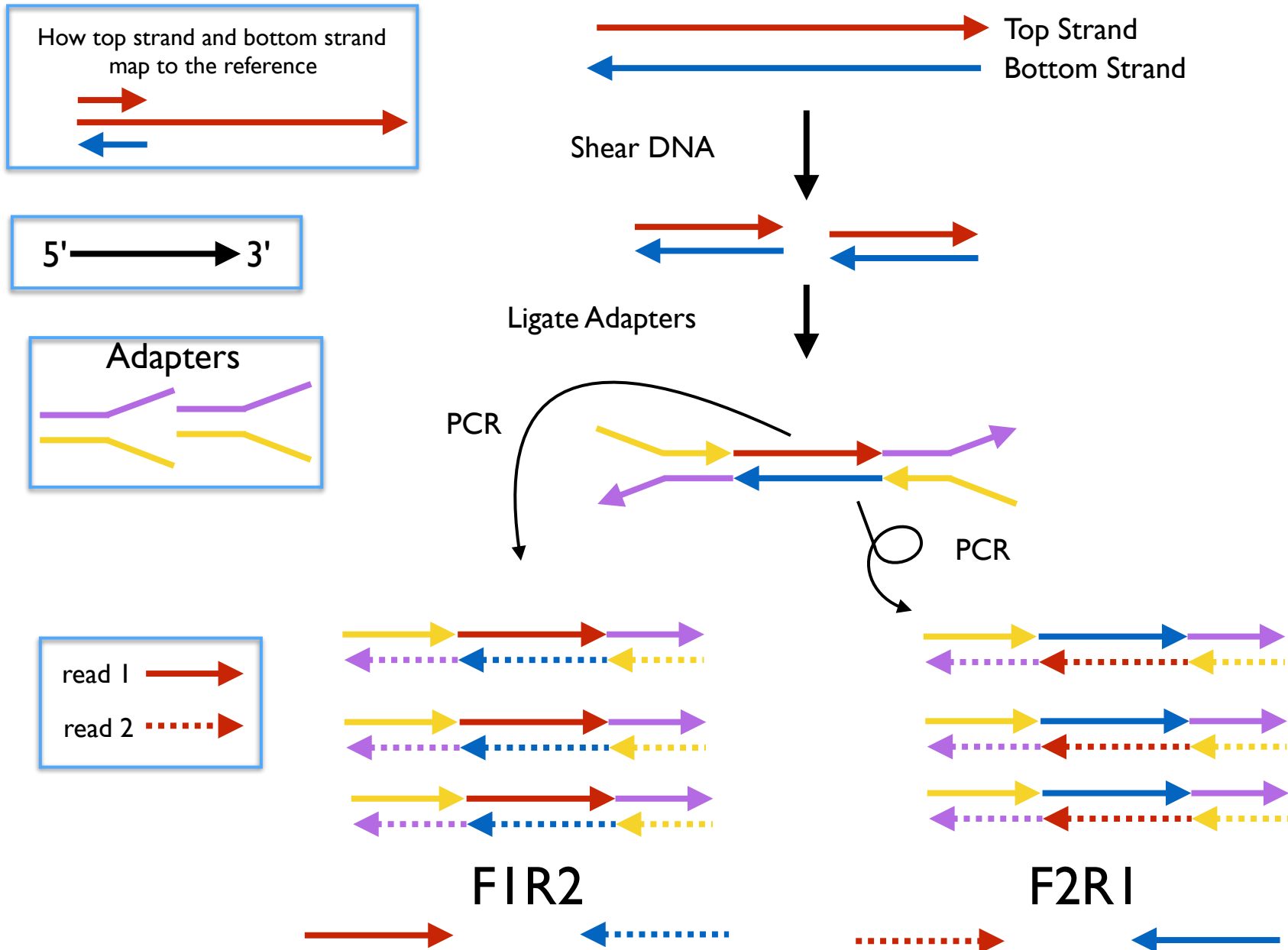


Read orientation tells us which strand (top or bottom) in the read came from original library

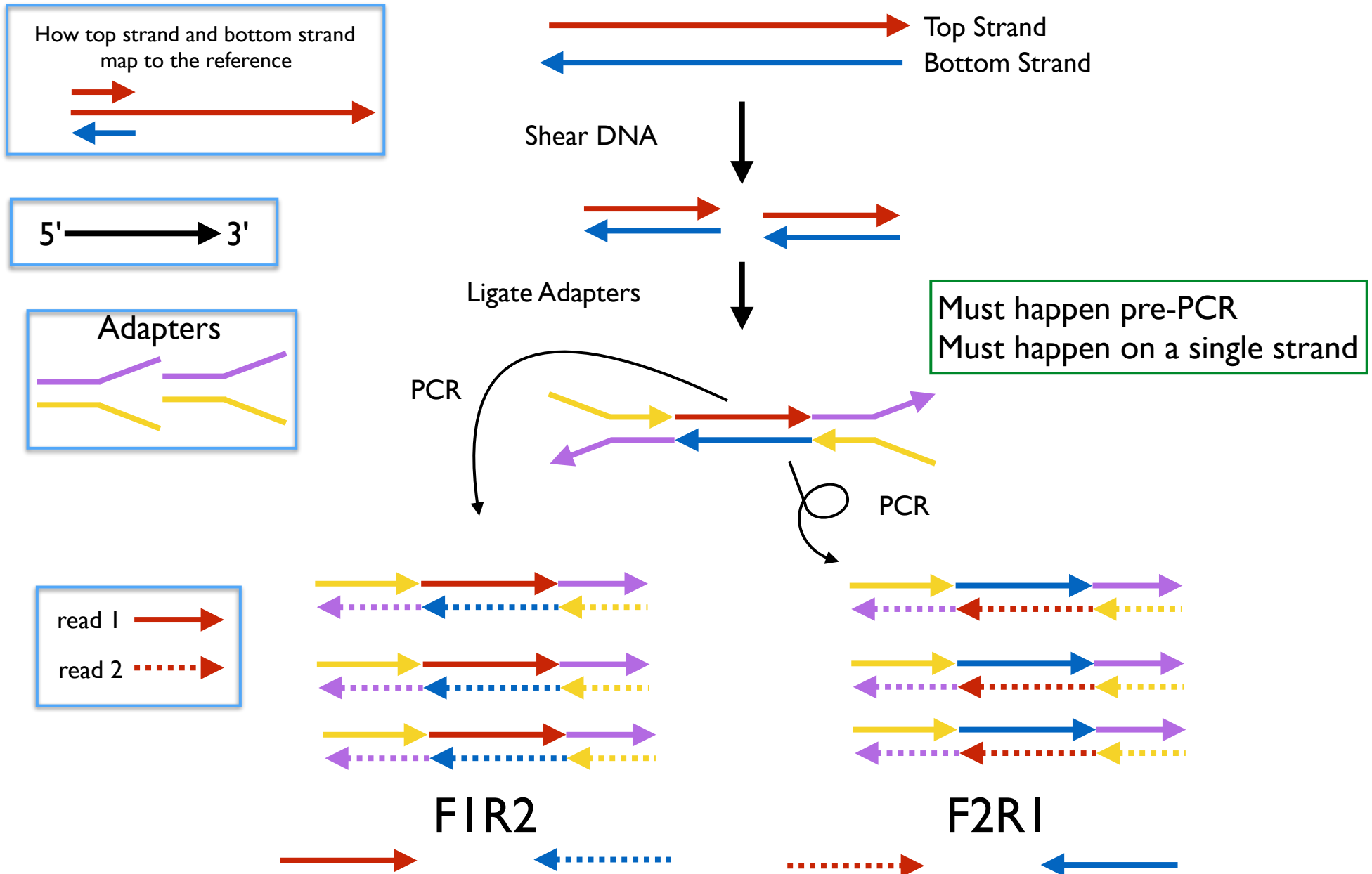
Read orientation artifact

- Alt bases are *only* in F1R2 (i.e. ones on the left)
- or *only* in F2R1 (ones on the right)
- When might we see an artifact like this?

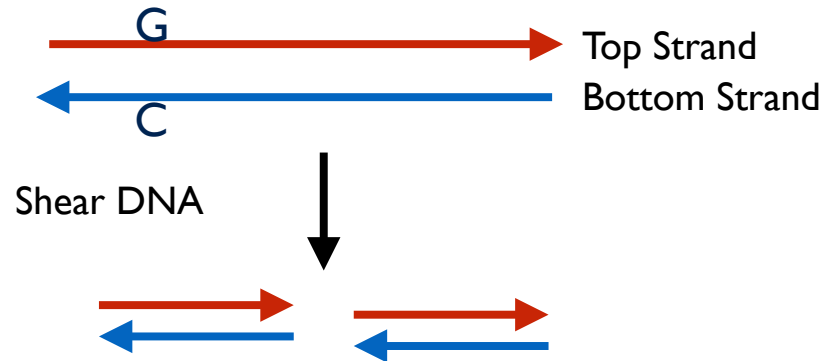
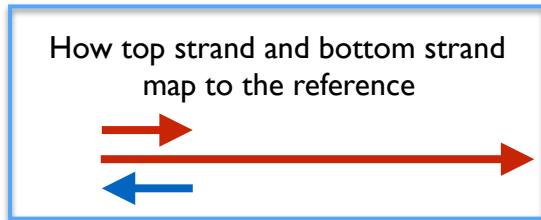
What is read orientation



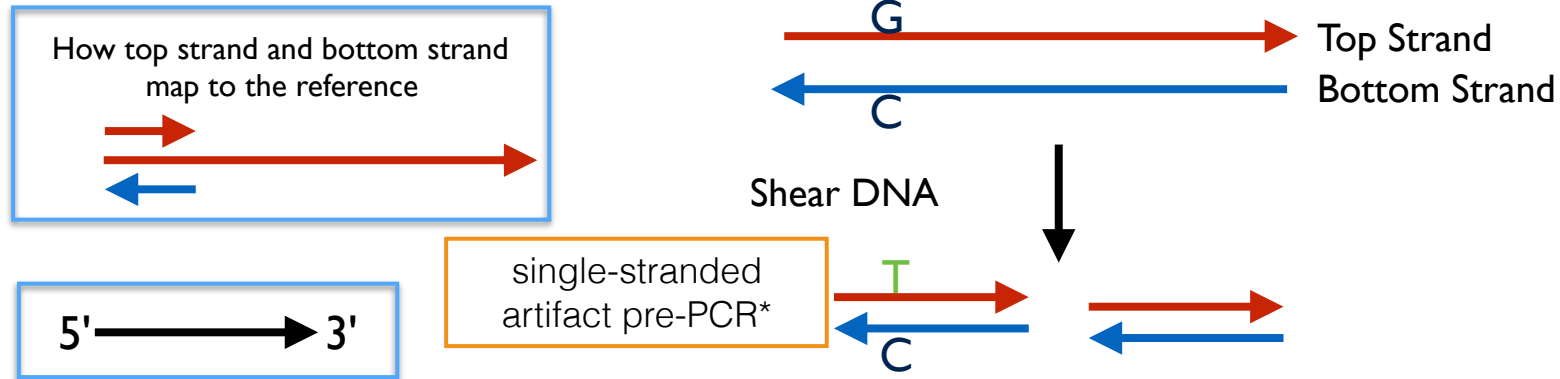
What is read orientation



Example: G -> T single-stranded artifact

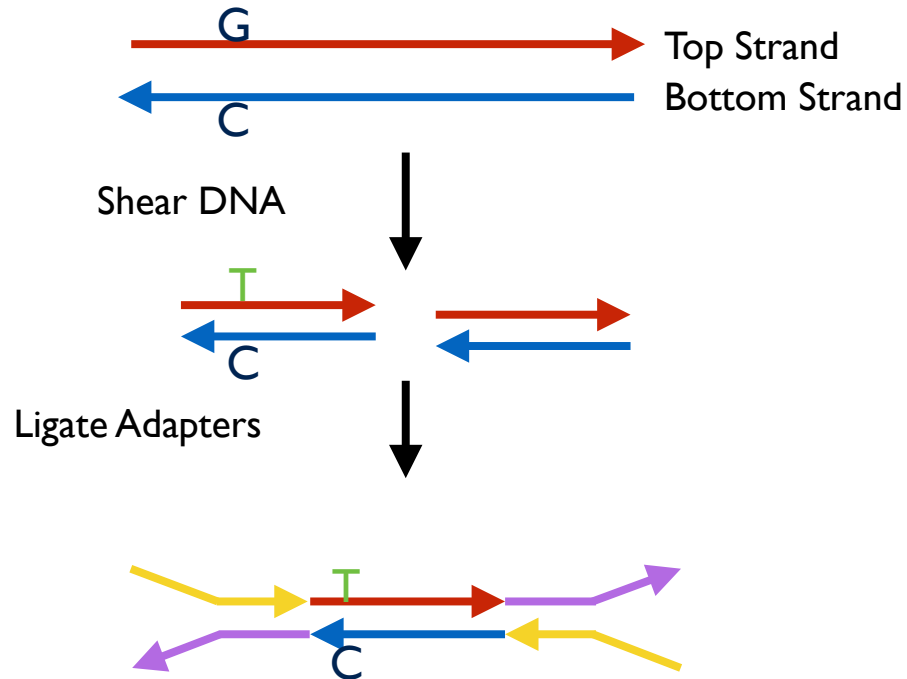
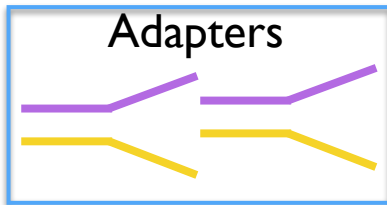
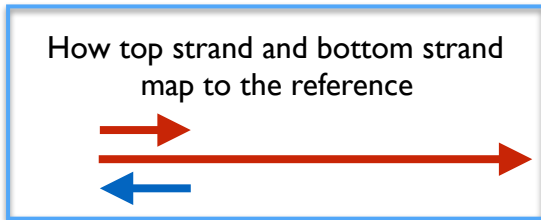


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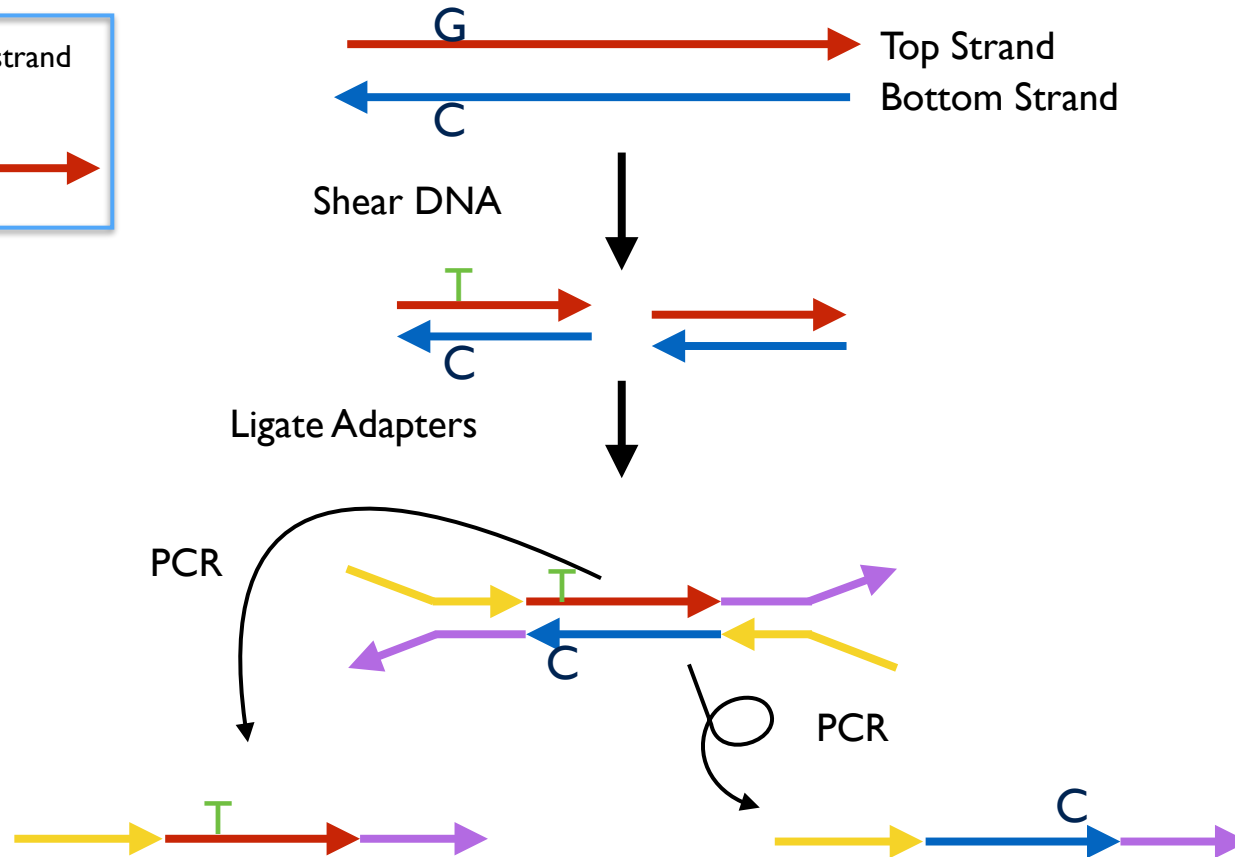
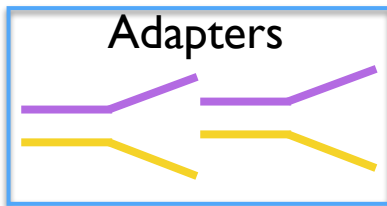
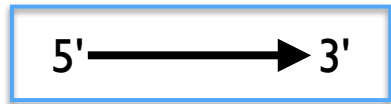
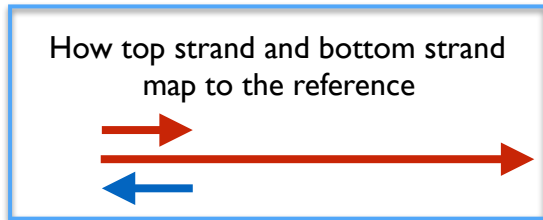


More precisely, G oxidizes and becomes oxo-G (G), which has high affinity for A i.e. G* acts like a T

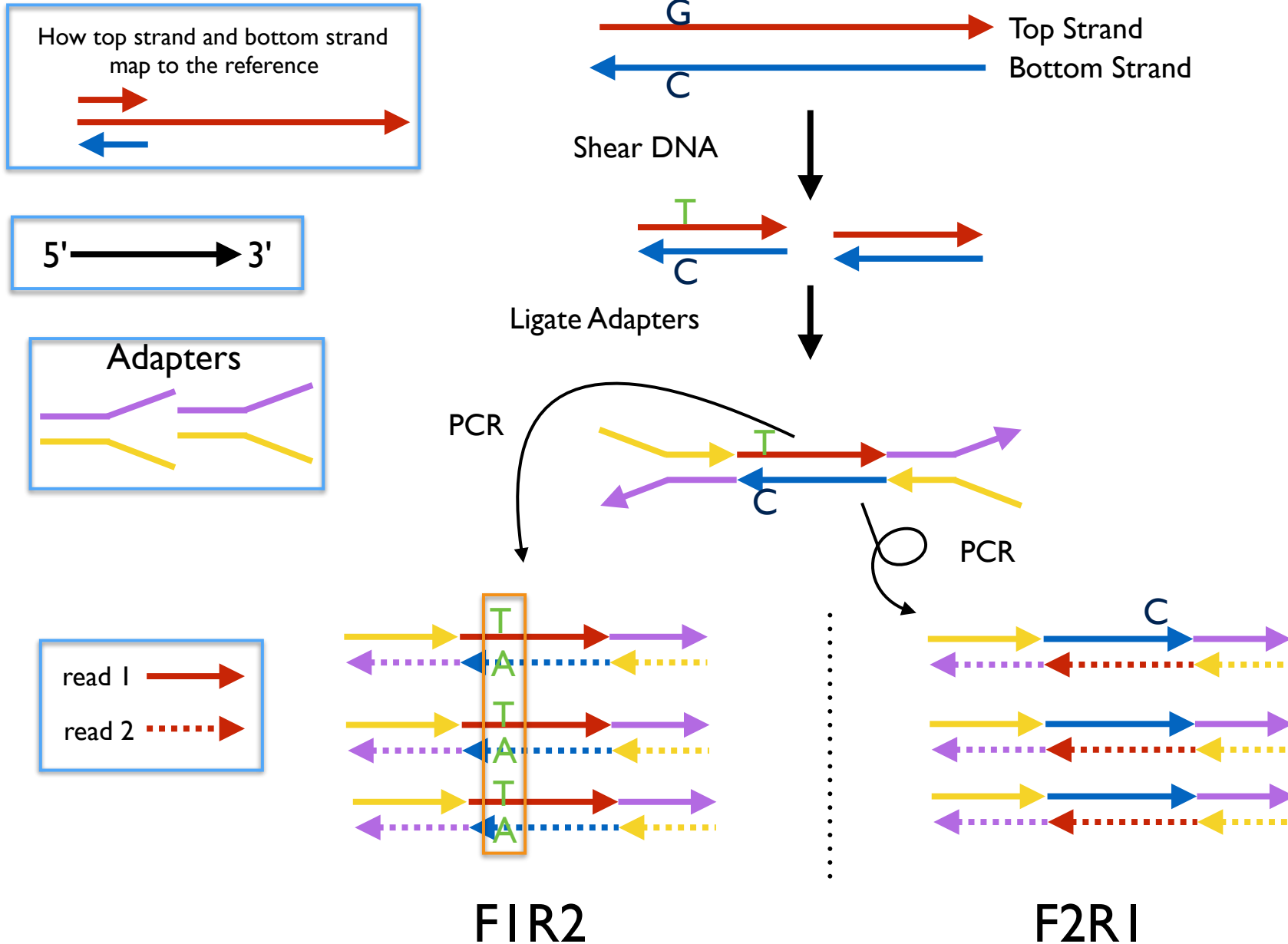
Example: G -> T single-stranded artifact



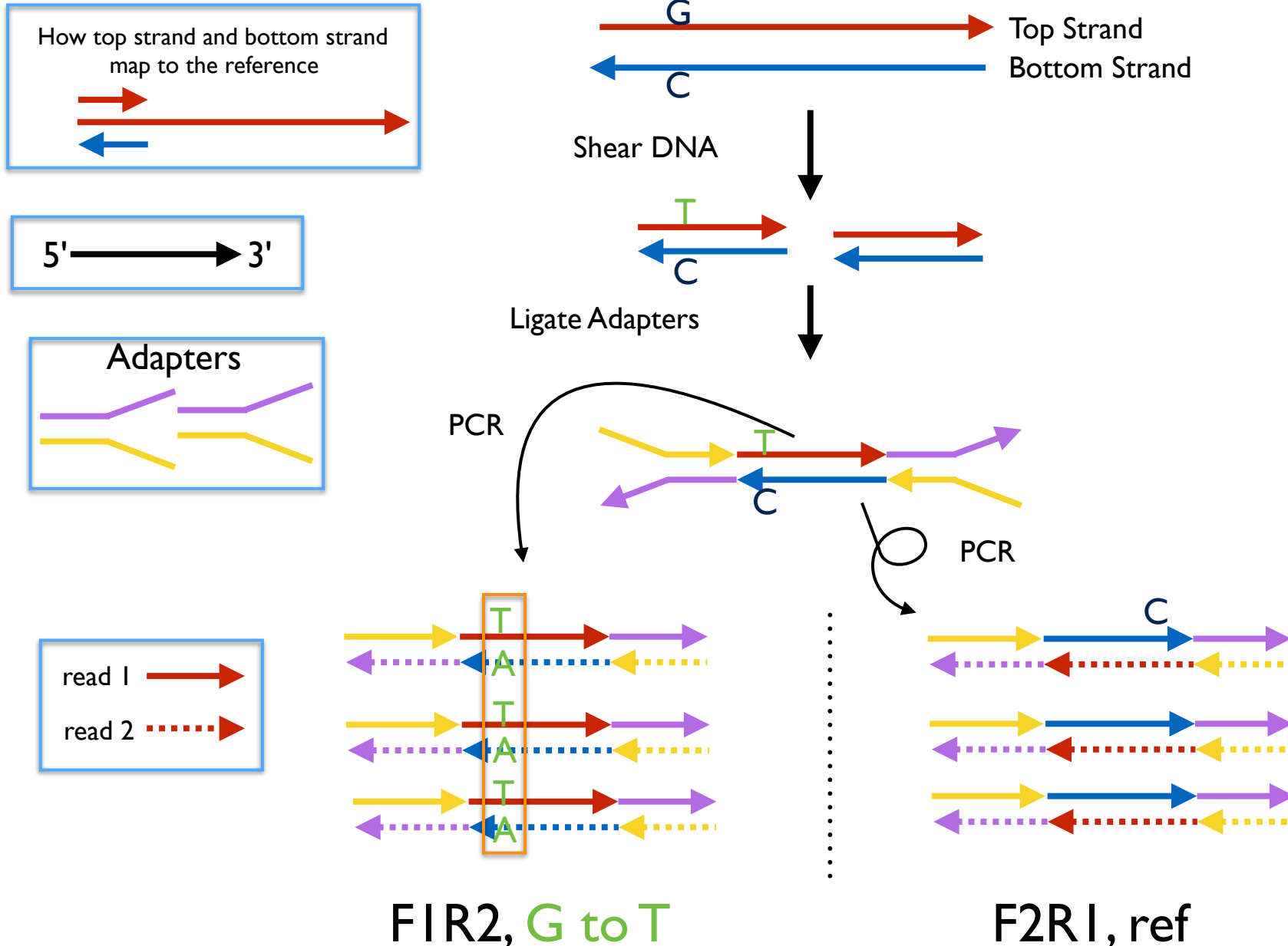
Example: G -> T single-stranded artifact



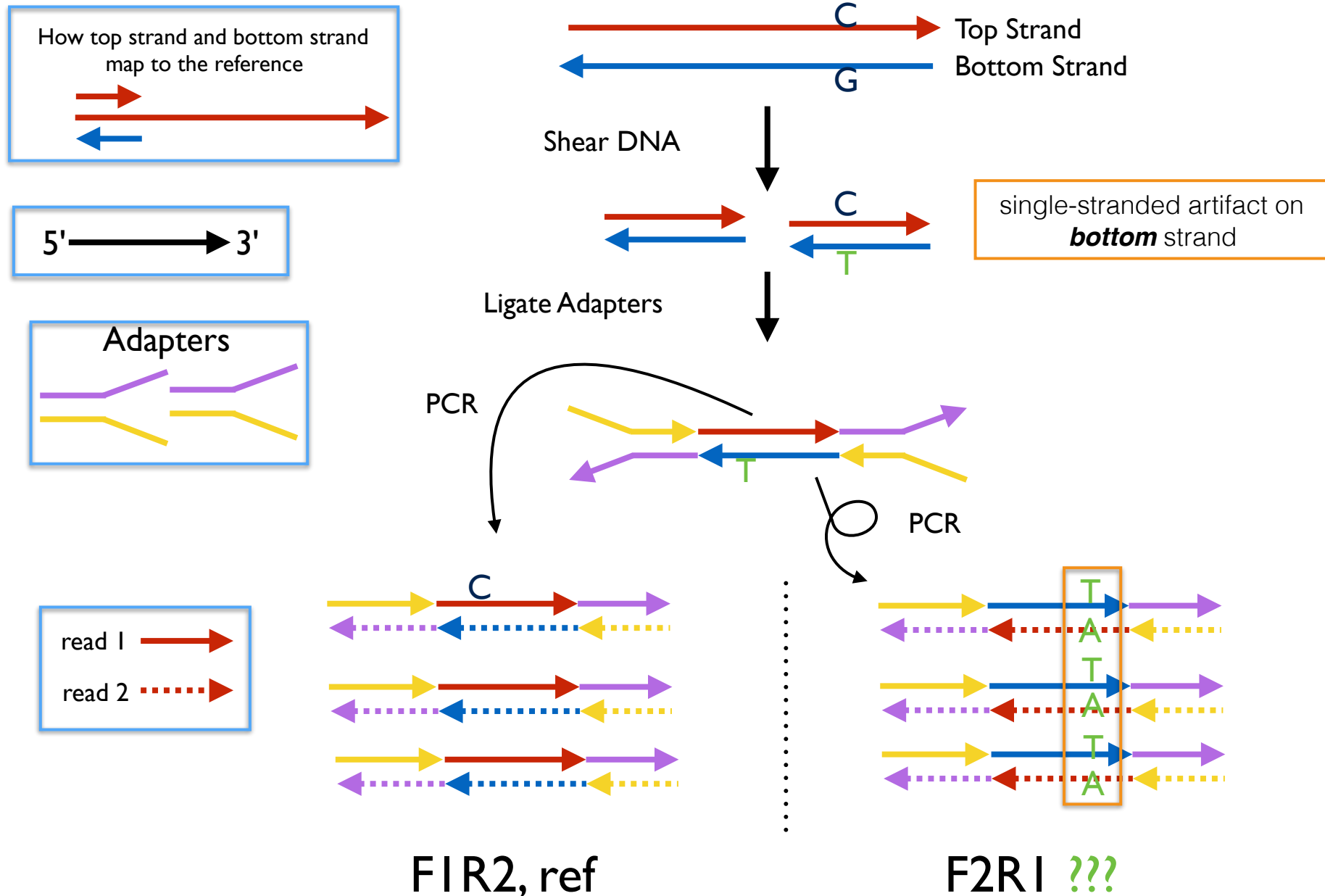
Example: G -> T single-stranded artifact



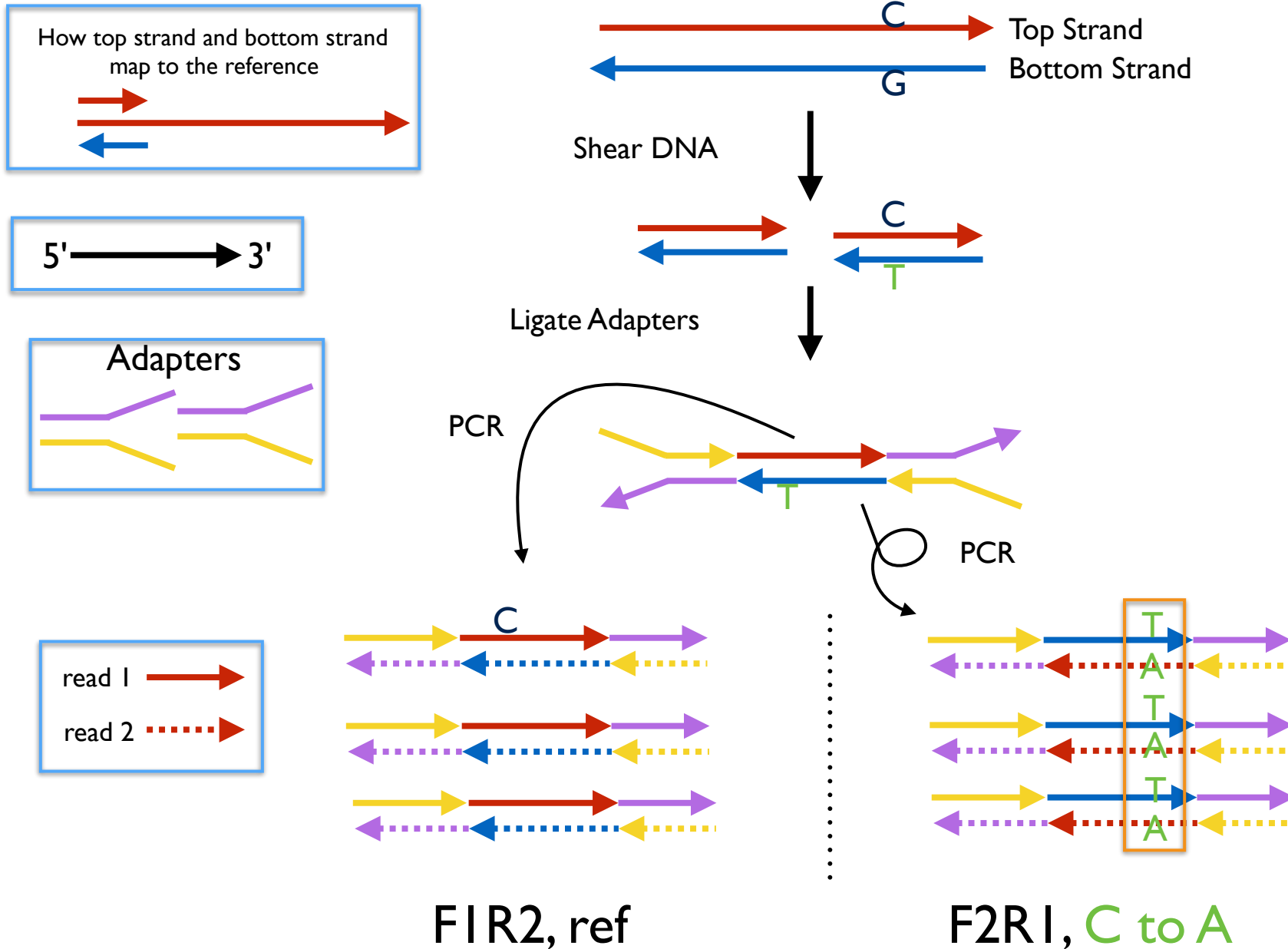
Example: G -> T single-stranded artifact



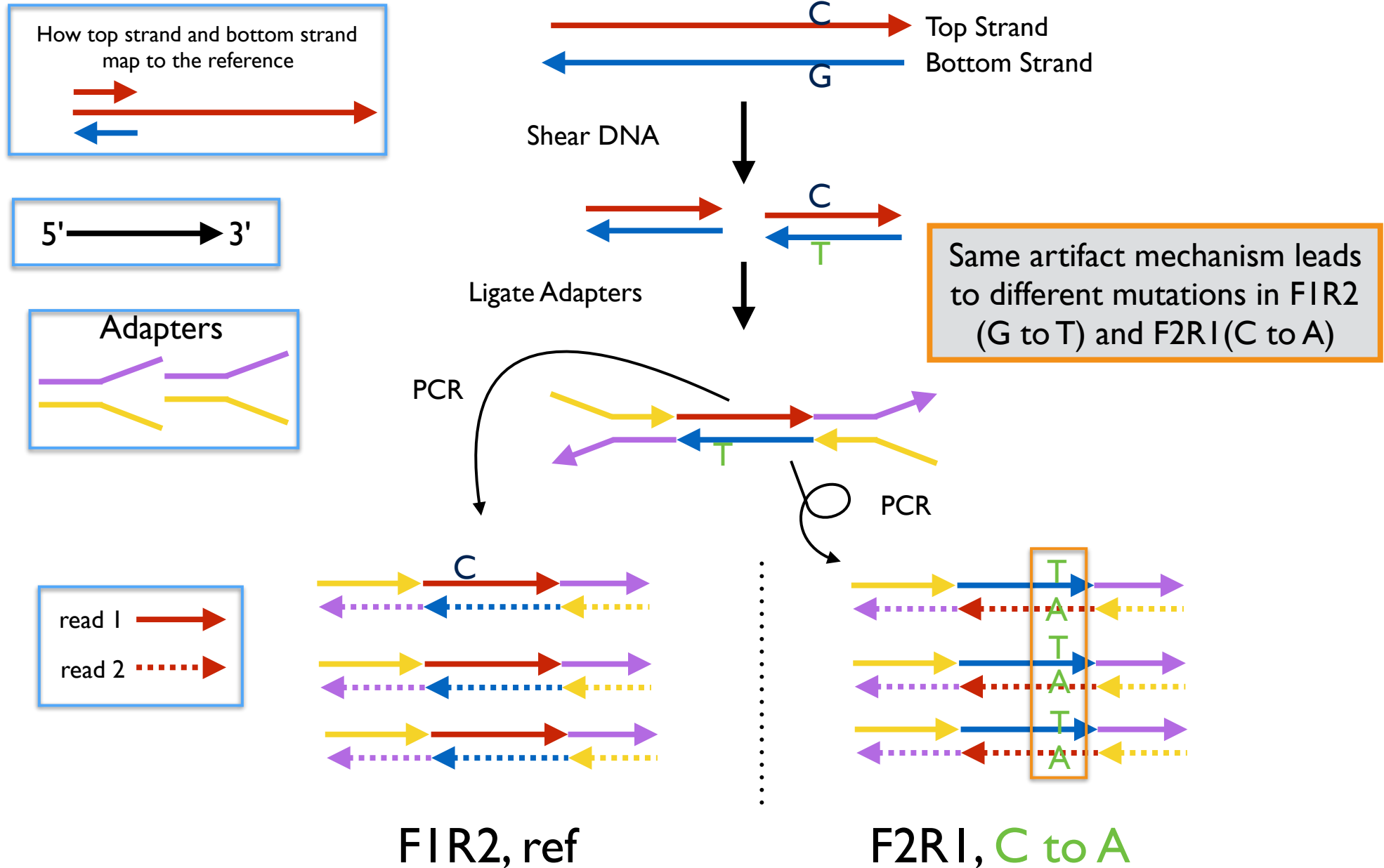
Example: G -> T single-stranded artifact



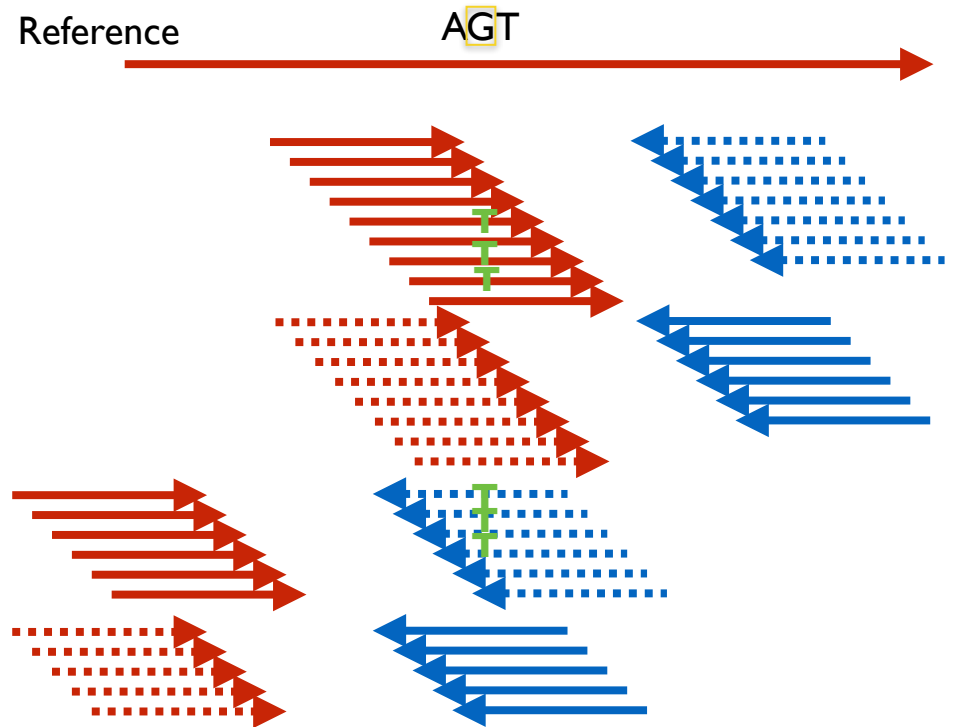
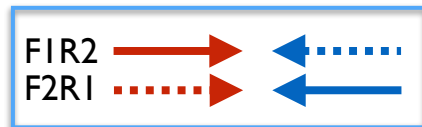
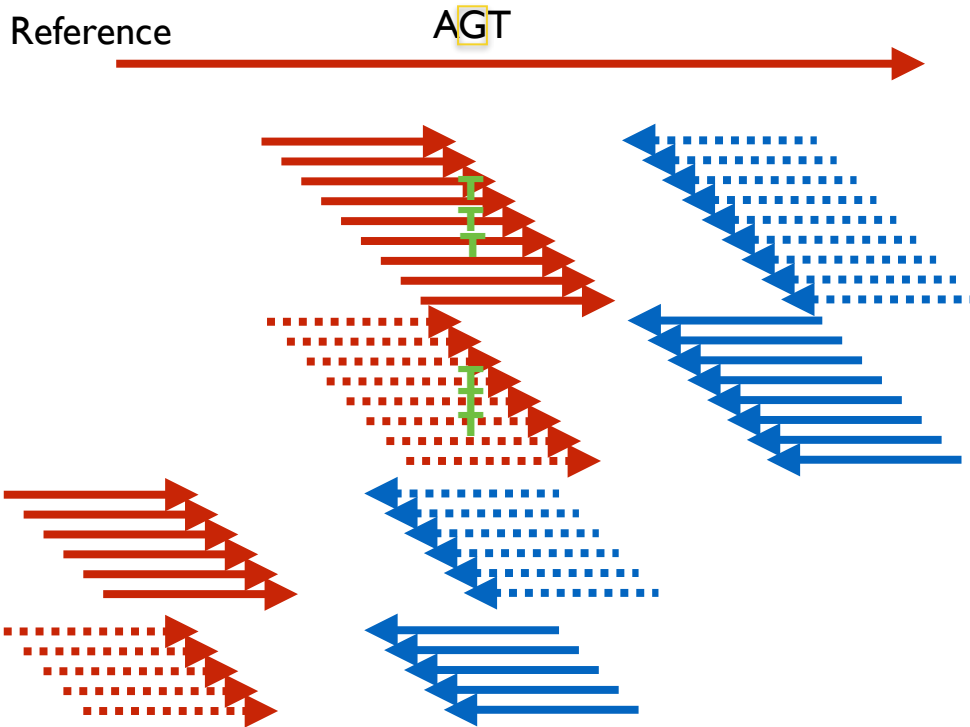
Example: G -> T single-stranded artifact



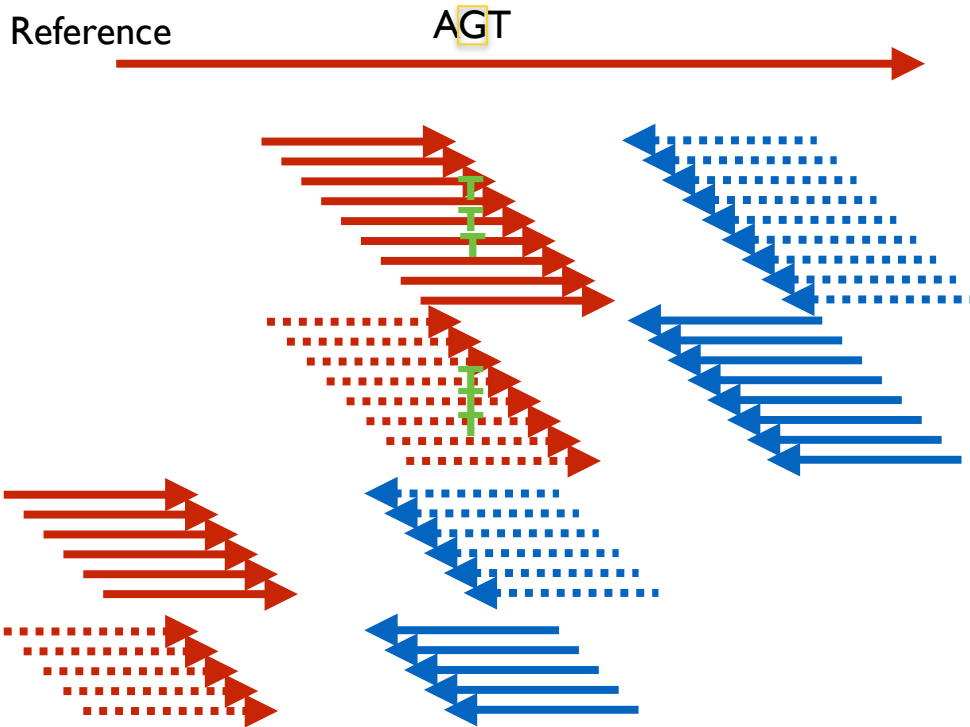
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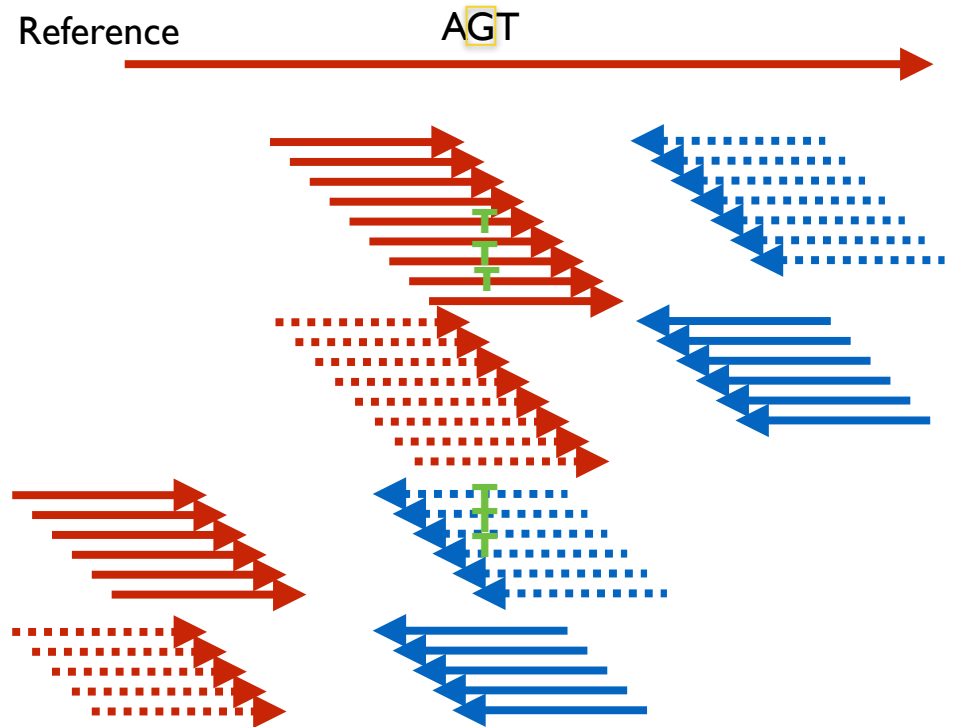
strand artifact != read orientation artifact



strand artifact != read orientation artifact



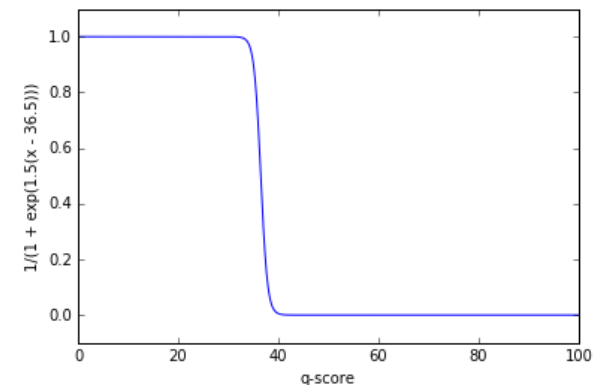
Strand Artifact



Read Orientation Artifact

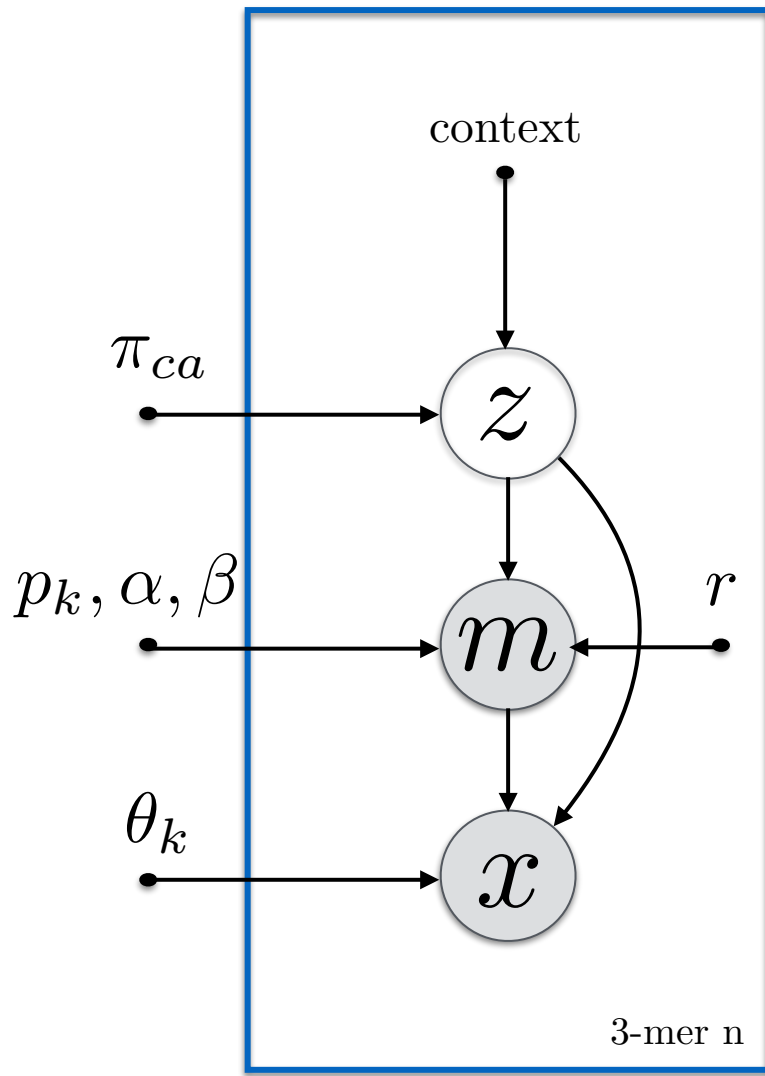
Existing filter

- Walk through bam, for each (ref context*, allele) pair, $Q = \text{qscore} = \text{phred}\left(\frac{\sum_{\text{sites}} \# \text{ Alt F1R2} - \# \text{ Alt F2R1}}{\sum_{\text{sites}} \text{Alt depth}}\right)$
- Higher Q = no artifact
- For each variant,
 - compute p-value with null hypothesis: $\# \text{ alt FIR2 reads} \sim \text{Binom}(\text{alt depth}, 0.96)$
 - False discovery = falsely call an artifact as non-artifact...
 - Reject non-artifact sites, Benjamini-Hochberg procedure to control FDR...or something
- Multiply # variants to filter by $\text{supressor} = \frac{1}{1 + e^{1.5(Q - 36.5)}}$
 - does not generalize
 - cannot detect rare events
- Must specify the artifact mode e.g. A -> T by hand
 - Requires manual inspection of collect sequencing metrics file
- Upshot - we need a new model



*ref context = 3-mer in reference

New Read Orientation Filter Model v4



$z \in \{\text{F1R2}_a, \text{F2R1}_a, \text{Hom Ref}, \text{Somatic Het}, \text{Germline Het Hom Var}\}$
 where $a \in \mathbb{A}$ and \mathbb{A} is a set of possible alt alleles in context c

$$z \sim \text{Categorical}(\pi_{ca})$$

$$r \equiv \text{depth}$$

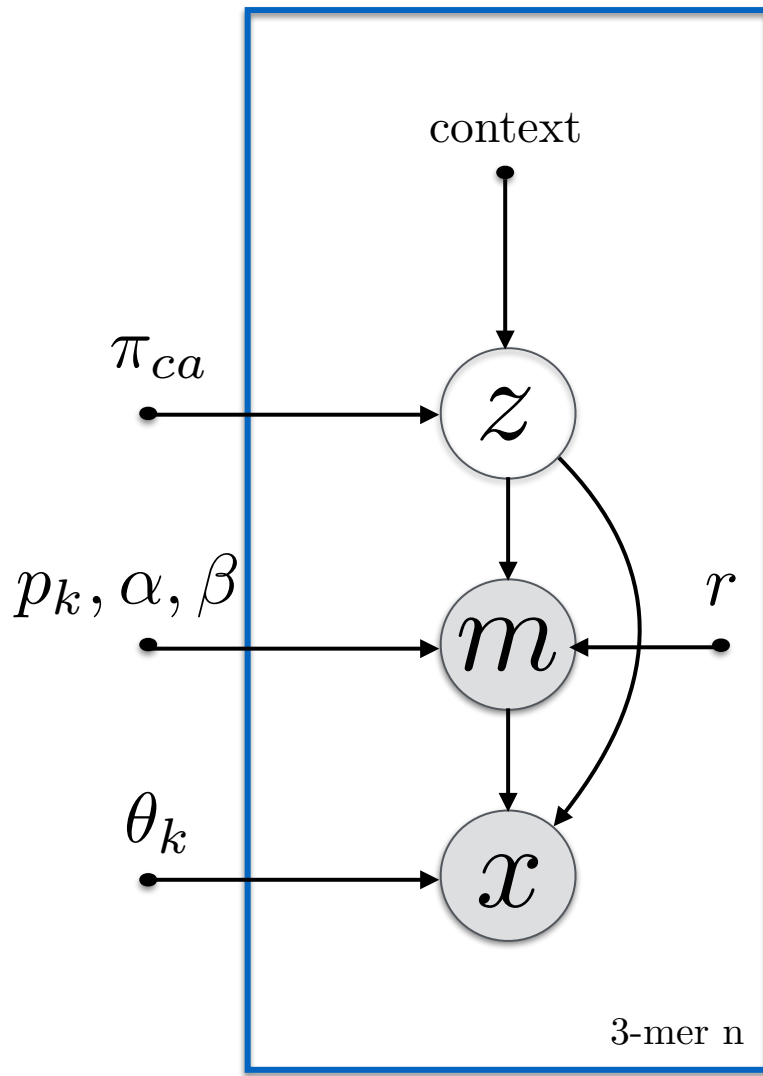
$$m \equiv \text{alt depth}$$

$$m|z \sim \begin{cases} \text{BetaBinom}(p_k, r, \alpha, \beta), & z = \text{Somatic Het} \\ \text{Binom}(p_k, r) & \text{otherwise} \end{cases}$$

$$x \equiv \text{alt F1R2 depth}$$

$$x|z, m \sim \text{Binom}(\theta_k, m)$$

New Read Orientation Filter Model v4



artifact

non artifact (F1R2 & F2R1 balanced)

$z \in \{\text{F1R2}_a, \text{F2R1}_a, \text{Hom Ref}, \text{Somatic Het}, \text{Germline Het Hom Var}\}$
 where $a \in \mathbb{A}$ and \mathbb{A} is a set of possible alt alleles in context c

$$z \sim \text{Categorical}(\pi_{ca})$$

$$r \equiv \text{depth}$$

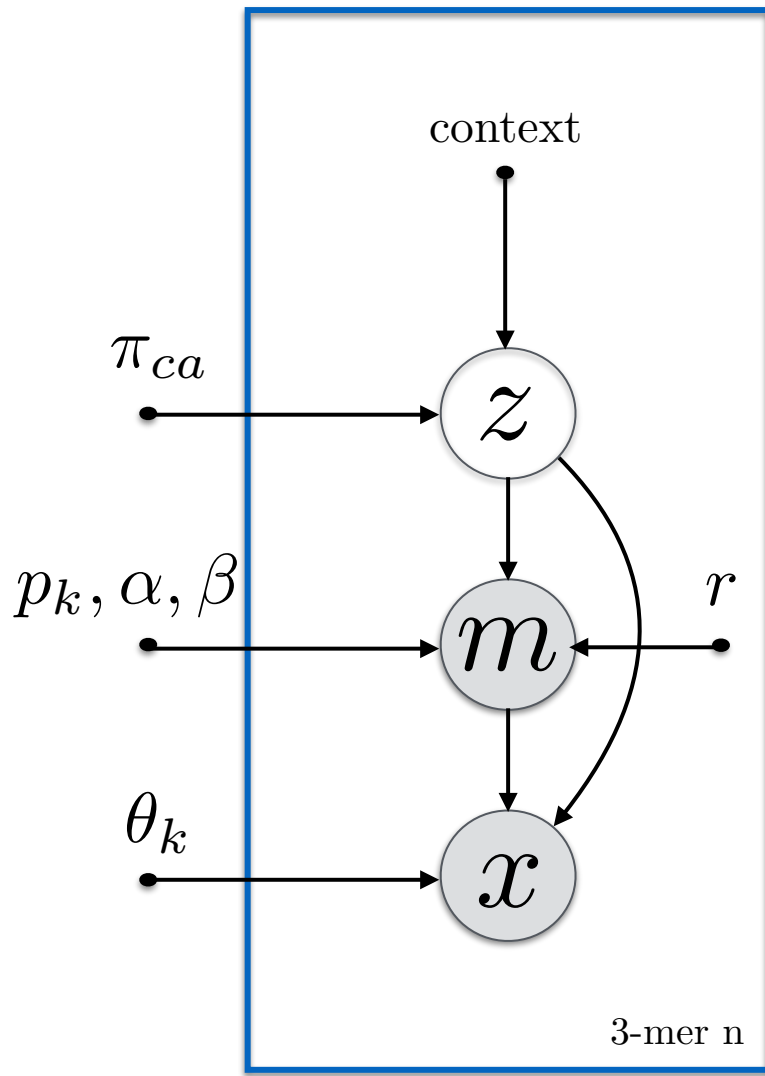
$$m \equiv \text{alt depth}$$

$$m|z \sim \begin{cases} \text{BetaBinom}(p_k, r, \alpha, \beta), & z = \text{Somatic Het} \\ \text{Binom}(p_k, r) & \text{otherwise} \end{cases}$$

$$x \equiv \text{alt F1R2 depth}$$

$$x|z, m \sim \text{Binom}(\theta_k, m)$$

New Read Orientation Filter Model v4



- Aware of the relative frequency of the artifact under each context
 - e.g. artifact in 1 in 1000 sites, 1 in 10 sites,
- can detect rare events
- No need to manually specify transitions you're looking for
- Simpler interpretation - posterior probabilities of z
- Replace CollectSequencingArtifact Metrics (pending Megan's approval)

New Read Orientation Filter Model v4

1. Learning Step (Name not ready for public announcement)
 1. estimate hyperparameters with EM
2. Inference Step (FilterMutectCalls)
 1. compute $p(z=F1R2|data)$, $p(z=F2R1|data)$

New Read Orientation Filter Model v5

1. Data Collection Step (Java, LocusWalker)

1. write out the design matrix to a file

2. Learning Step

1. estimate hyperparameters with EM, powered by PyMC

3. Inference Step (FilterMutectCalls/Java)

1. for each variant, compute the posterior probabilities of z , and filter