

NGS READ PROCESSING

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FASTQ files

- Text file with four lines per read

1. Label
2. Sequence
3. +
4. Quals

```
@M141:79:749142:1:1101:14941:1421 1:N:0:GTTATCCGTACA  
TACGTAGGTGGCAAGCGTTATCCGGAATTATTGGGCGTAAAGCGCGAGTAGGCGGTTTTTA  
+  
=====--+55,55@@@EEA6>A.6>C7C>BFGG=AEC5+@EF=ED7+5CEF=ACDC55AE)5
```

- Format not fully standardized
 - Different conventions for representing Q scores as letters
 - Software may have different max & min Q scores
 - Typical is Q2 to Q40

Quality (Phred) scores

- Integer Q2 .. Q40
- Represents P_{error} , probability base is wrong

Q40: $P_{error} = 0.0001$	99.99% good
Q30: $P_{error} = 0.001$	99.9% good
Q20: $P_{error} = 0.01$	99% good
Q10: $P_{error} = 0.1$	10% wrong
Q3: $P_{error} = 0.5$	50% wrong
Q2: $P_{error} = 0.66$	<u>66% wrong!!</u>

Quality filtering

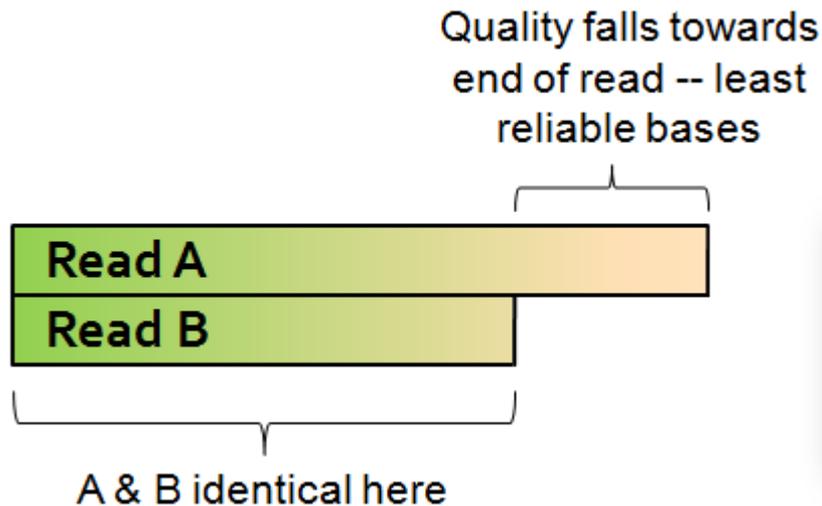
- Discard poor-quality data
- Poor quality = high probability of error(s)
 - low Q scores
- Genomics can mask out low-Q positions
 - e.g. for SNP-calling

Quality filtering

- Amplicon sequencing different scenario
 - Need pair-wise comparisons for most analysis
 - pairs of reads, or reads & database
 - to calculate identity or determine if sequences identical
 - Masked / ambiguous positions (Ns) problematic
 - Variable length (e.g. truncated at low Q) also problematic
- OTU clustering
 - "Harmful" reads $>3\%$ errors create spurious OTUs
 - High diversity in harmful reads
 - Many spurious OTUs even if harmful reads small fraction

Truncating at low Q is bad idea

- Read quality often falls towards end of read
- Popular (but bad!) to truncate when Q low



Do A and B have identical sequences?

If **Yes**, dubious tail gets high abundance

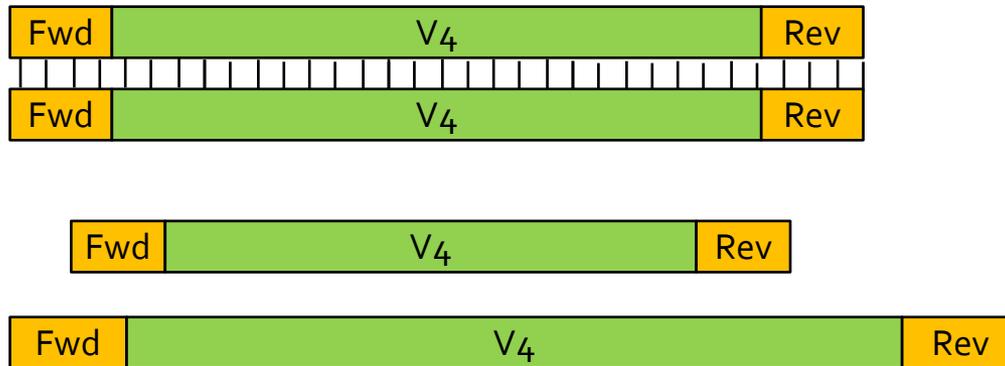
If **No**, good prefix gets low abundance

Length trimming

- Similar/identical reads should be globally alignable with few/no terminal gaps
- Comparisons unambiguous
 - Cannot have A identical (or >97% similar) to prefix of B
- Unpaired reads: truncate to fixed length
 - Important for 454
 - Often not needed for Illumina
 - Sometimes trim low-quality tails

Global trimming

- Full-length amplicons with varying length ok
 - e.g. overlapping paired reads
 - trim to primers ok
 - no terminal gaps when same / closely related



Quality filtering methods

✓? Minimum Q

- Ok if Q is large, e.g. $Q \geq 20$ ($P_{error} = 1\%$)
- Ok if don't truncate -- keep or discard

✗ Average Q, maybe over sliding window

- Conceptual mistake -- averaging logarithms!?
- Errors dominated by small Qs

✗ QIIME filter

- Truncate (👎) read if >3 consecutive bases with $Q \leq 3$
- $Q=3$ means $P_{error} = 50\%$
- Allows reads with **many** errors!

Quality filtering methods

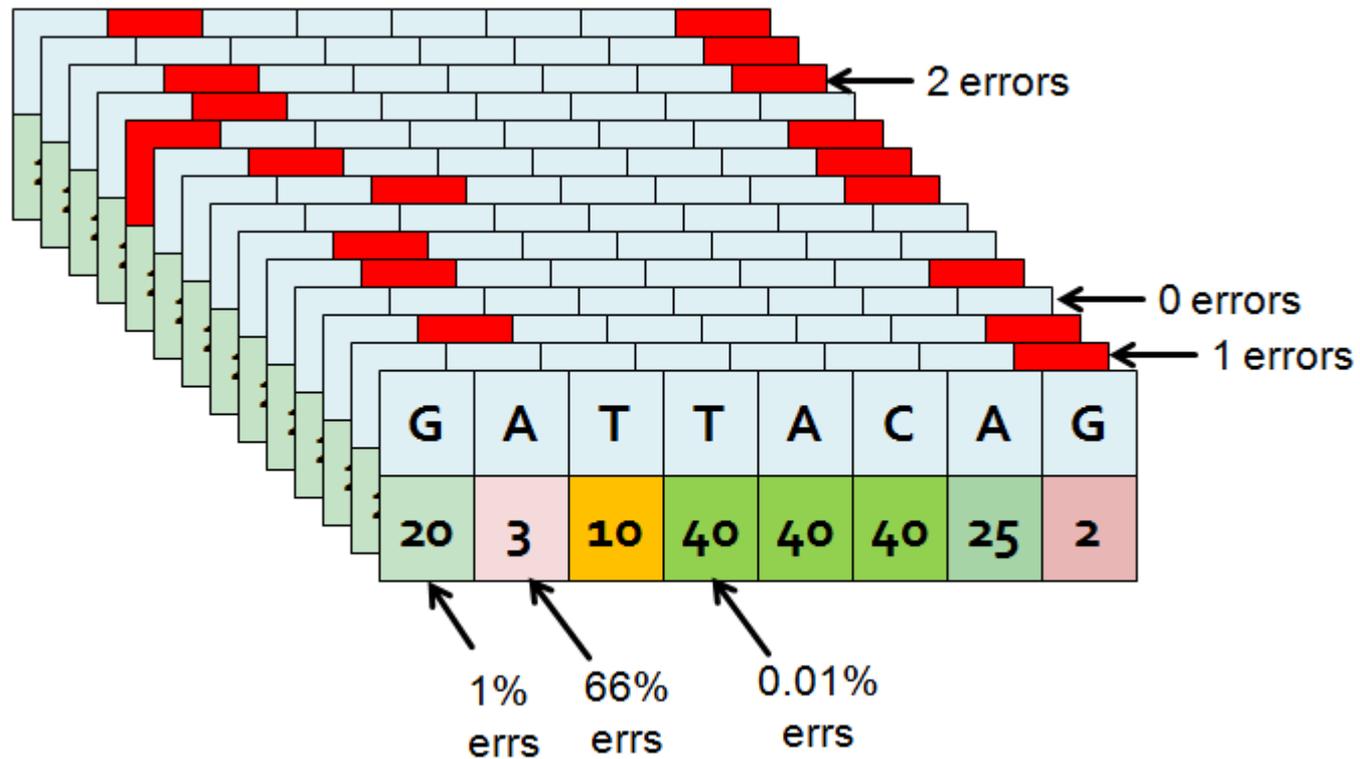
✘ PANDAseq method

- t = geometric mean of $P_{correct}$ along read ≥ 0.6
- $P_{error} = 0.4$
- Much too high, allows reads with many errors
- Better with higher t , but not as good as expected errors

✓ Expected error filtering

G	A	T	T	A	C	A	G
20	3	10	40	40	40	25	2

Expected errors



Expected errors

- Expected errors (E) in a read
- E = mean over large set with random errors according to Q scores
 - real-valued (because it's an average)
 - always > 0
 - can be < 1

Expected errors

- Surprisingly easy to calculate E
Sum the error probabilities
 $E = \text{sum } P_{\text{error}}$
- Most probable number of errors E^*
 $E^* = \text{largest integer } \leq E$
 $= \text{floor}(E)$
- Proofs in Edgar & Flyvbjerg (2015).

Expected error filter

- Discard reads with $E > 1$
 - Keep reads with $E^* = 0$
 - Most probable number of errors = zero
- Typical performance on MiSeq 2x250 V4
 - keeps 75%+ of the reads
 - 2/3 of filtered reads are correct (zero errors)
 - 1/3 have one or more bad bases

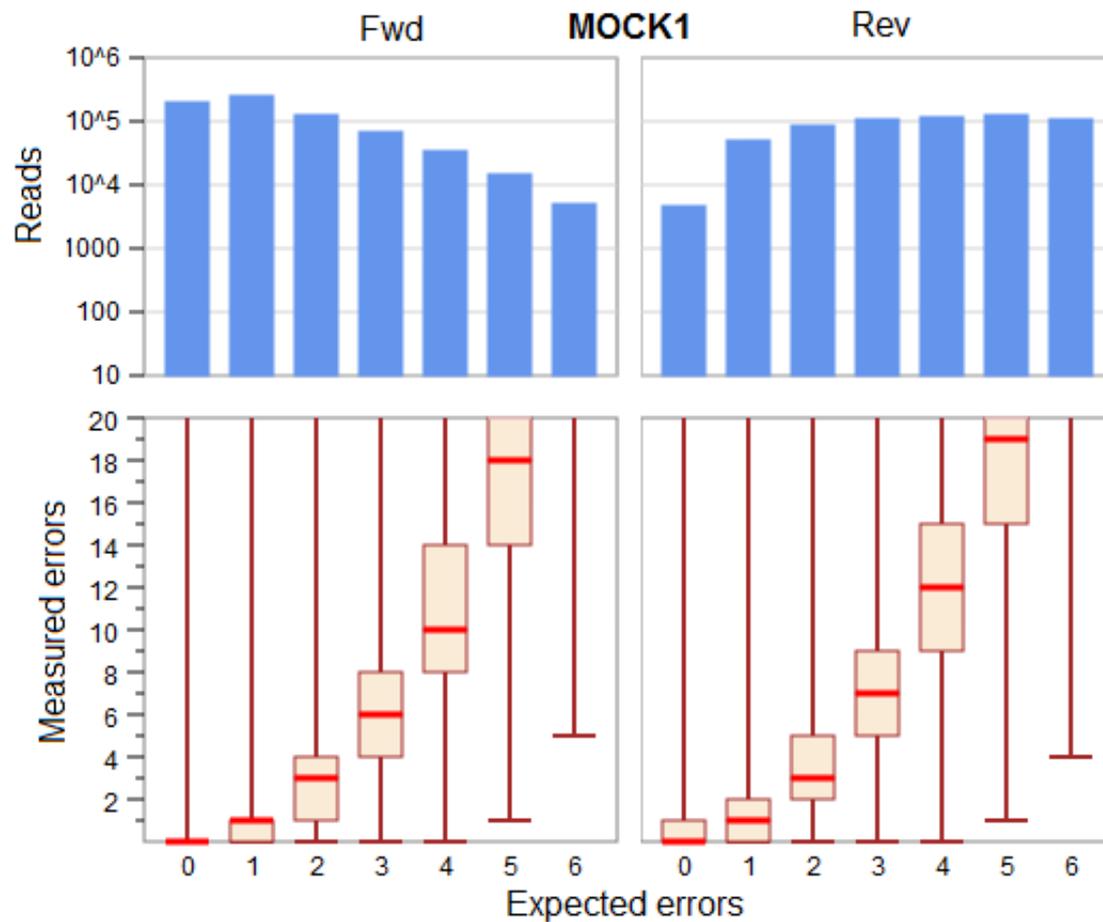
Expected error filter

- Works well if Q scores are accurate
- Illumina Q scores are pretty good
- 454 not so good
 - filtering not so effective
 - expected error filter still best method
- Max $E=1$ suggested default
 - Not a requirement! (note for comparative validation)
 - Larger E for less stringent filtering (more spurious OTUs)
 - Smaller E for very stringent filtering

Expected error filtering

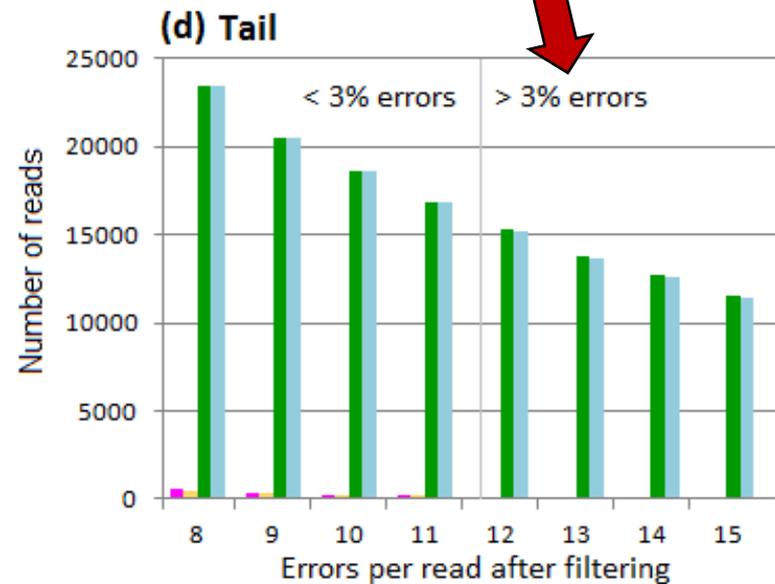
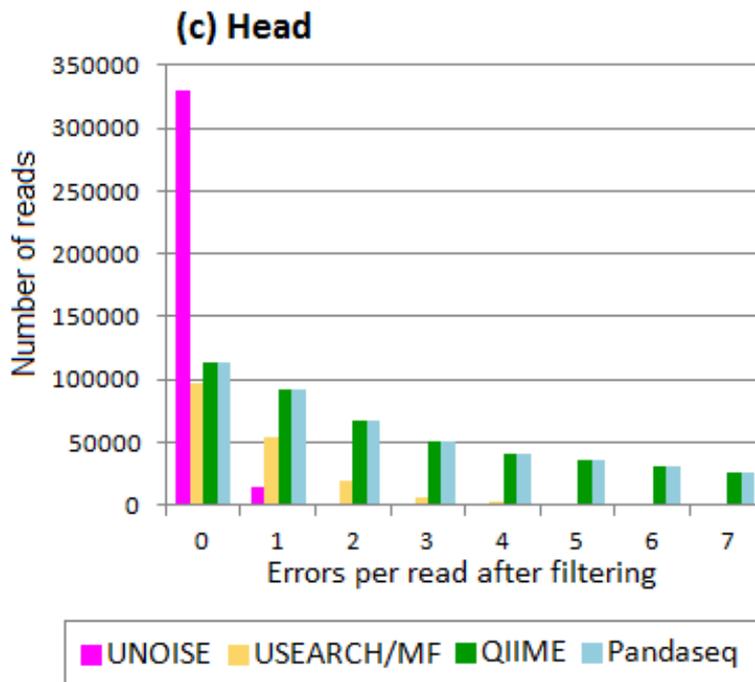
- Critics: allege too stringent
 - high cost in sensitivity, diversity
- Reads are not lost!
 - Most filtered reads map to OTUs after clustering
 - Filtering is critically important to suppress spurious OTUs
- High sensitivity to rare species not possible
 - Contaminants, cross-talk...
 - Limit of resolution abundance $> \sim 0.5\%$ of reads

Expected vs. measured errors

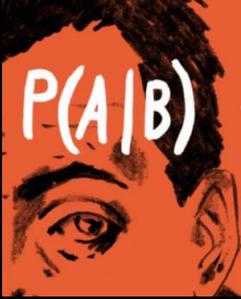


Quality filter performance

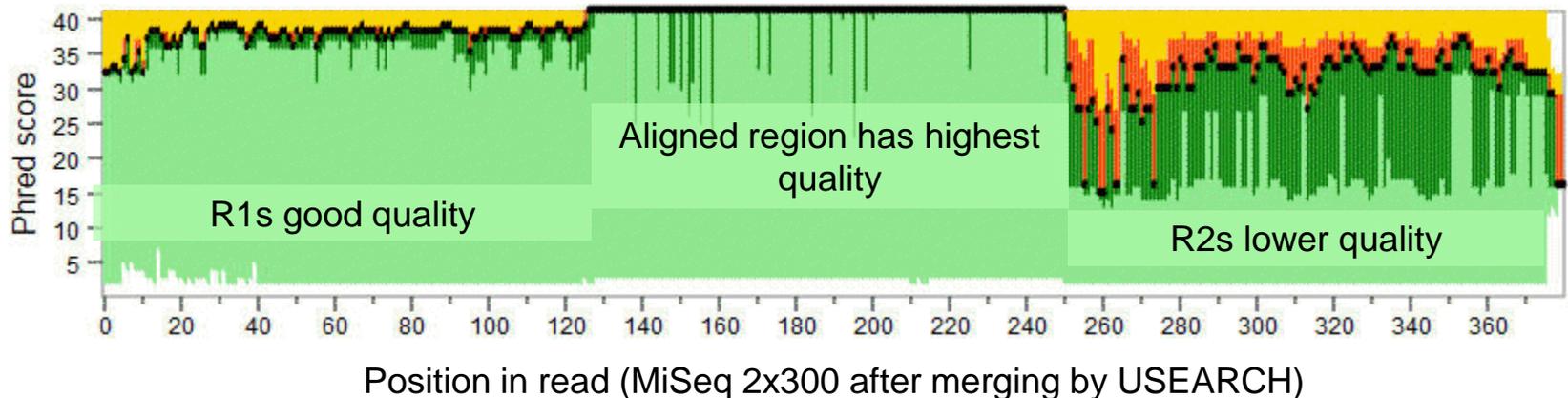
QIIME and PANDAsq filters leave tens of thousands of reads with >3% errors, thousands of spurious OTUs



Paired read merging


$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

- Two observations of each base in overlap
- Should increase/decrease Q if match/mismatch
- Use Bayes' Theorem to get posterior P_{error}
 - Correct equations in Edgar & Flyvbjerg (2015)
 - Previous papers got this wrong!

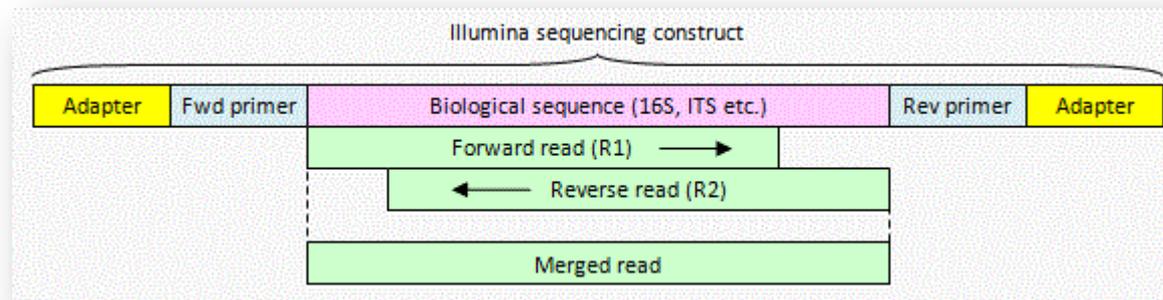


Paired read assemblers

Program	Mean match Q error	Mean mismatch Q error	Random merge test
BBmerge v8.82	-6.1	-0.5	Pass
COPE v1.1.2	-15.3	13.6 (25% wrong sign)	Pass
FLASH v1.2.11	-8.8	-0.4	Pass
fastq-join Download 21 Nov 2015	-8.7	-0.3	Pass
mothur v.1.36.1 Make.contigs	<i>(Uses PANDAseq method to calculate Qs, but recommends not to use for quality filtering).</i>		Fail (100% assembled) <i>(May not be a problem if mothur SOPs are followed).</i>
PANDAseq v2.8	-21.6 (73% wrong sign)	-11.6 (20% wrong sign)	Fail (~70% assembled)
PEAR v0.9.5	-1.3	15.11 (27% wrong sign)	Pass
SeqPrep Dated 6 Jan 2015	7.3	-0.6	Pass
USEARCH v8.1 fastq_mergepairs	0	0	Pass

Do we need full overlap?

- V₄ is ~250nt
- 2x250 PE reads give full overlap
 - Better error correction?
- Accurate OTUs with UPARSE on **R2s only!**
- Longer amplicons ok, e.g. V₃-V₄ (400nt)
 - better resolution



Dereplication

- Find the unique sequences in the reads
 - and their abundances
- Abundance is a very useful signal
 - Most abundant sequences almost certainly correct
 - unless low-Q truncated
 - Errors increasingly common at lower abundances
- Pool reads from all samples
 - Strongest abundance signal

Singletons

- Abundance = 1
- Random errors usually singletons
 - Not usually reproduced by chance
- Systematic errors may have $ab. > 1$
 - Polymerase errors & chimeras (amplified by PCR)
 - Sequencing error usually pretty random

Discard singletons

- After filtering, many reads with $>3\%$ errors
 - Sequencer error
 - Polymerase copying errors
 - Chimeras
 - Most of these are singletons
- Discard singletons before clustering
 - Necessary to minimize spurious OTUs
 - Most singletons map to OTUs after clustering, not lost!

Discard singletons

- Critics: allege high cost in sensitivity, diversity
- Effect on sensitivity minimal / meaningless
 - By definition, found once in one sample!
 - Ecologically irrelevant (or not possible to interpret)
- Sensitivity is < 100% with singletons
 - Sampling effects, e.g. rare species missed
 - Primer mismatches ("universal" = ~80% - 90%)
 - Some / many rare species missing regardless
- Diversity metrics like Chao1 **nonsense** for 16S

Delete primer-binding sequences

- PCR tends to substitute mismatches
- Not needed with many Illumina protocols
 - 16S / ITS primer-binding sequence not in read