

# When Good Experiments Go Bad

Chris Miller

“

However improbable we regard this event, or any of the steps which it involves, given enough time it will almost certainly happen at least once.

--George Wald

“Anything that can go wrong, will go wrong

--Murphy

“  
Shit happens.

--Forrest Gump

# Case #1

- Exome sequencing – Glioblastoma Tumor/Normal pairs
- Alignment, somatic variant calling, filtering

```
$ wc -l H_RL-01-0*/snvs.indels.annotated
```

```
159 H_RL-01-0203-1412449/snvs.indels.annotated
```

```
219 H_RL-01-0216-1412454/snvs.indels.annotated
```

```
10583 H_RL-01-0334-1412447/snvs.indels.annotated
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34-1412447/snvs.indels.annotated

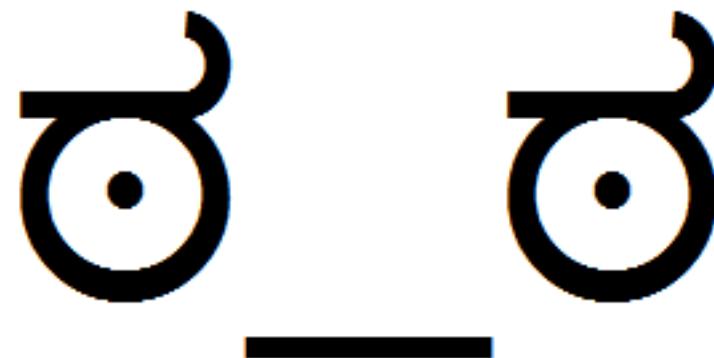
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- How many of these occur at known dbSNP sites?

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- How many of these occur at known dbSNP sites?  
~85%

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- What is their VAF?

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- How many of these occur at known dbSNP sites?

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- What is their VAF?

~50% or 100%

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**dx: SAMPLE SWAP**

# Damage Control

- Check other samples in the cohort
  - May not be resolvable!

S1 Tumor vs S1 Normal #####

S2 Tumor vs S2 Normal #####

S1 Tumor vs S2 Normal ##

S2 Tumor vs S1 Normal ##

are the tumors or normals swapped?

- Often need more information to resolve (RNAseq? Cytogenetics?)
- Check other lanes/indices on the same machine/batch
- Often, the resolution is to drop the samples

# Case #2

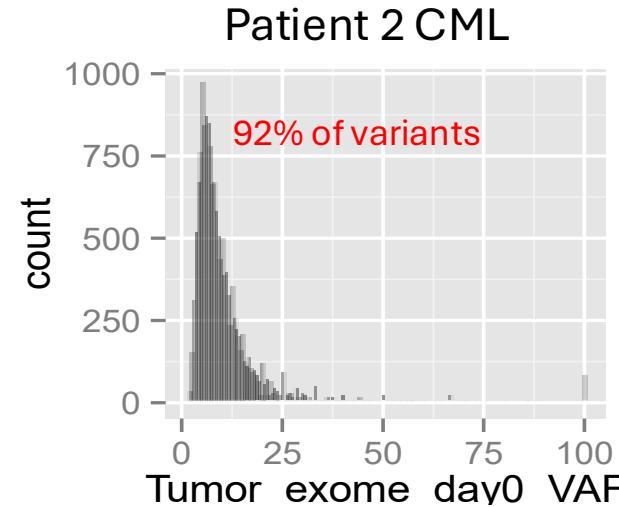
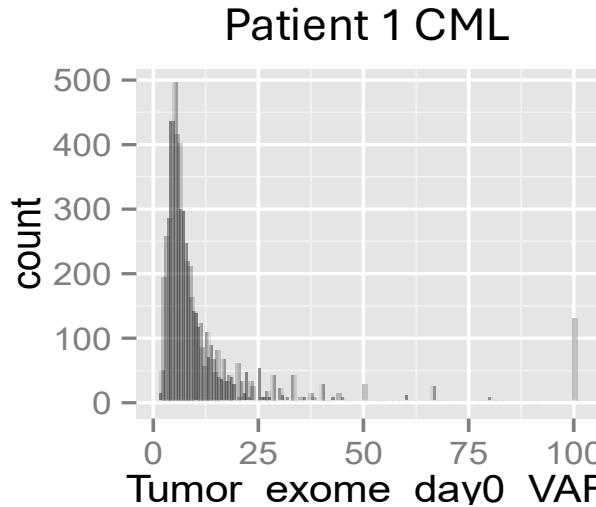
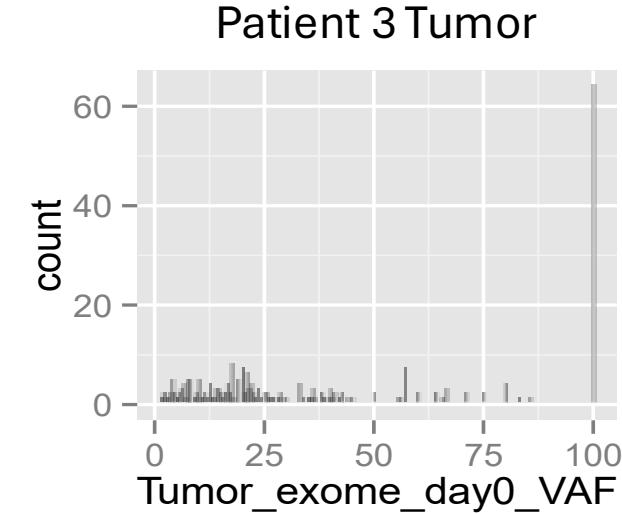
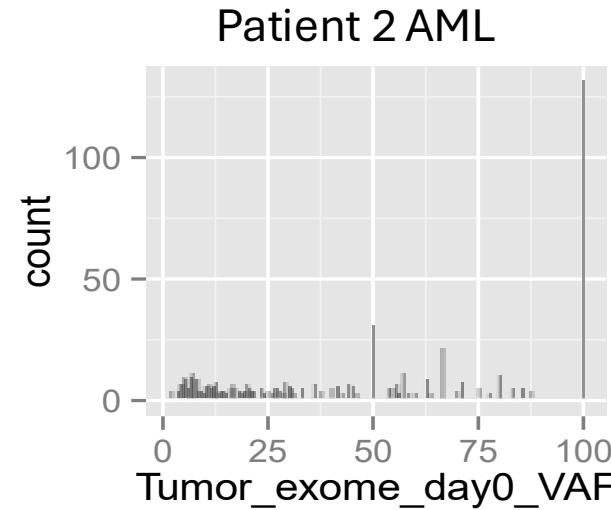
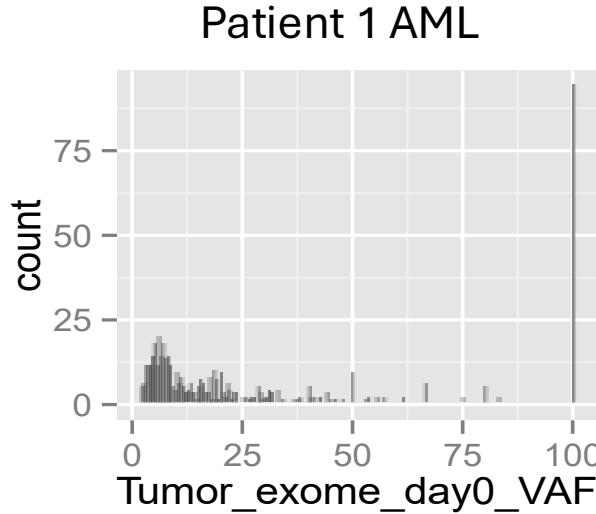
- 2 projects, 3 patients
  - Patient 1 – Normal, CML, AML
  - Patient 2 – Normal, CML, AML
  - Patient 3 – Normal, Tumor
- Exome Sequencing
  - 1 lane of HiSeq2500 - 79-99X mean depth

Variant	Patient 1 CML	Patient 1 AML	Patient 2 CML	Patient 2 AML
SNVs	14430	669	5276	87
Indels	255	115	55	17

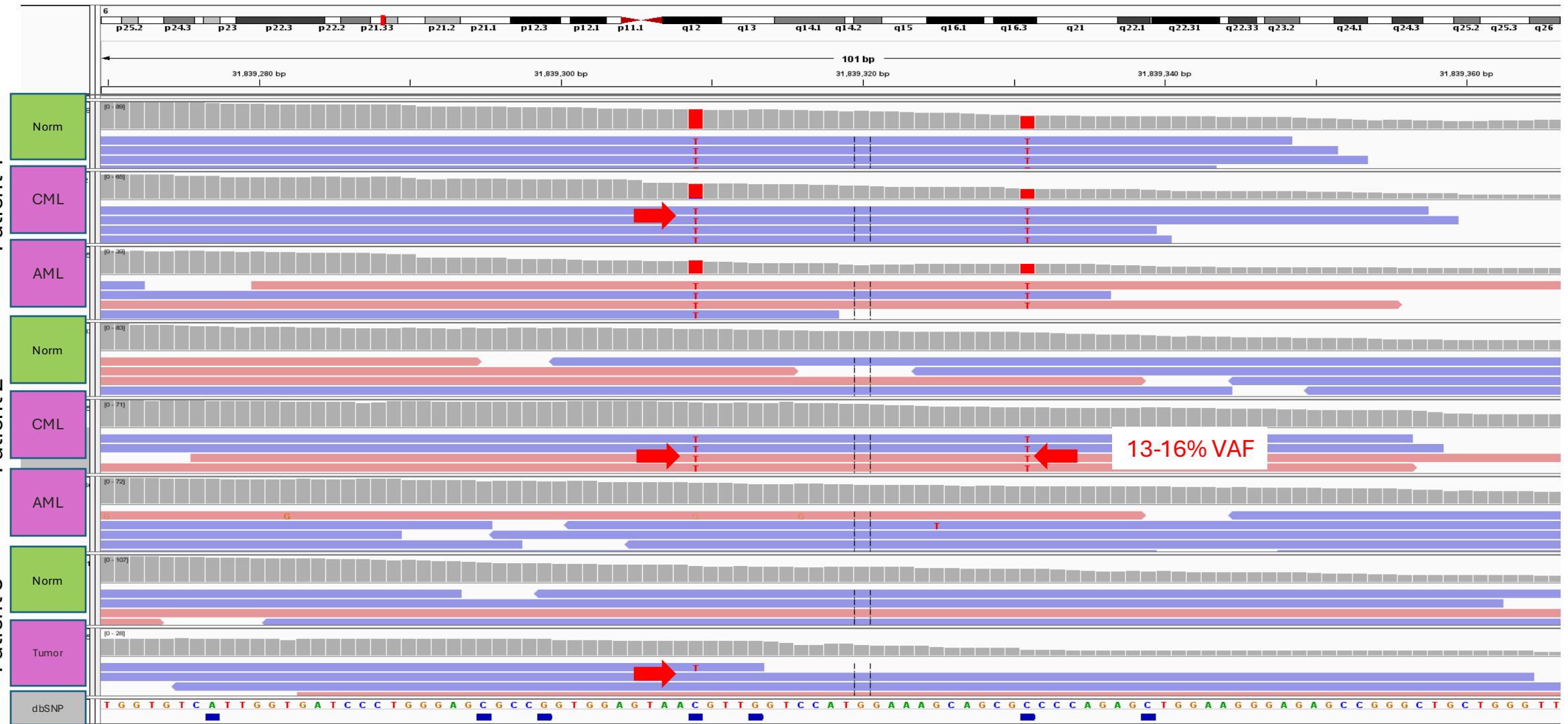
# How many SNVs are common variants?

Variant	Patient 1 CML	Patient 1 AML	Patient 2 CML	Patient 2 AML
SNVs	14430	669	5276	87
Indels	255	115	55	17
in dbSNP	>90%	<5%	>90%	<5%

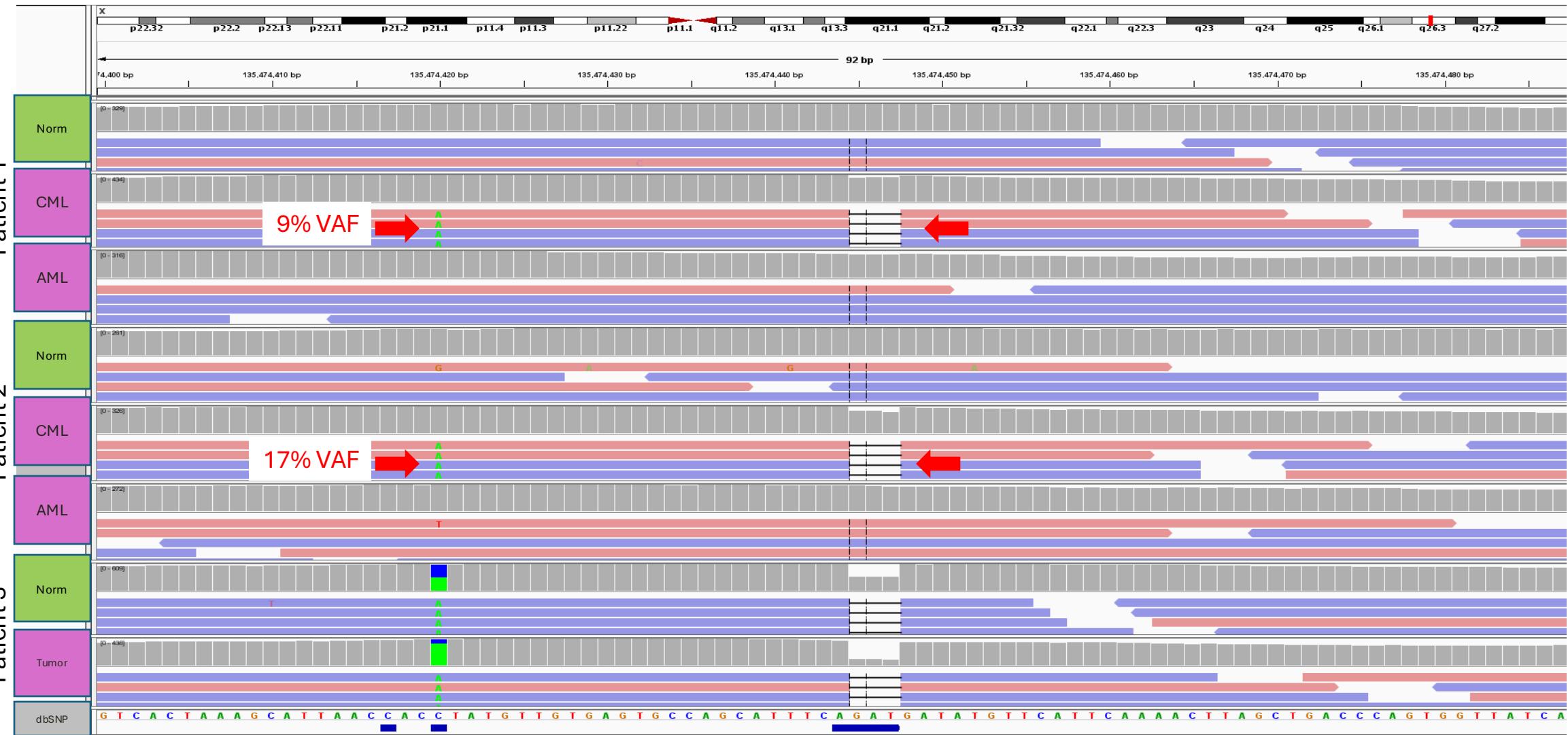
# Coding variants with an rsID (dbSNP)



# IGV inspection of variants/reads



# IGV inspection of variants/reads

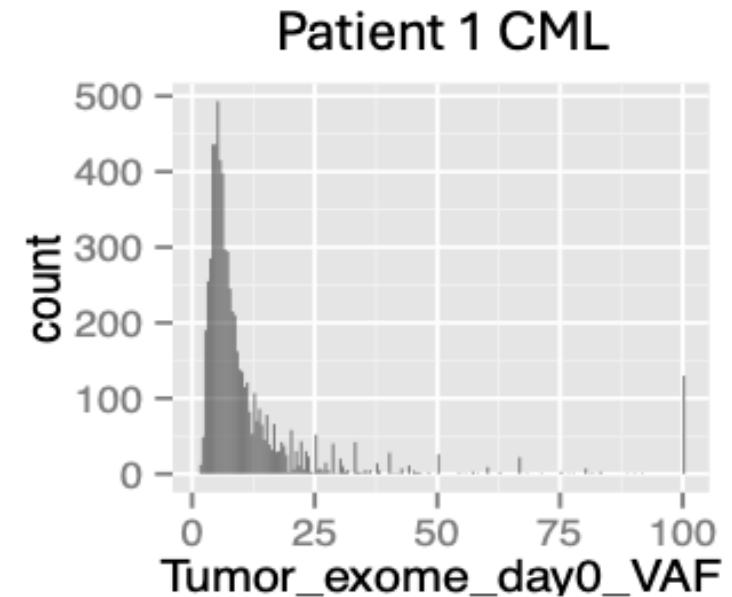


- How many of these occur at known dbSNP sites?  
most
- What is their VAF?  
NOT ~50% or 100%

**dx: SAMPLE CONTAMINATION**

# Damage Control

- Check other samples in the cohort to figure out source
  - May not be resolvable! (what if it's from someone else's samples?)
- If it's low-level enough, could apply filters
  - Only keep VAFs >30%
  - Downside: you may miss real events!
- Best solution is to make new libraries from the original source tissue



# Case #3

- Single-cell RNA sequencing data
- Transcriptome alignment
  - we expect high level: 90%+
- Our data had ~10% alignment

This example is good, ours was not!

## Mapping ?

Reads Mapped to Genome	100.0%
Reads Mapped Confidently to Genome	21.4%
Reads Mapped Confidently to Intergenic Regions	2.6%
Reads Mapped Confidently to Intronic Regions	12.5%
Reads Mapped Confidently to Exonic Regions	6.3%
Reads Mapped Confidently to Transcriptome	16.3%
Reads Mapped Antisense to Gene	2.0%

# Case #3

- Checked the kit - 3' vs 5' (matched)
- Checked the data – blatted a read at random

Human BLAT Search

BLAT Search Genome

Genome:  Search all genomes

Assembly:

Dec. 2013 (GRCh38/hg38)

Query type:

BLAT's guess

Sort output:

query,score

Output type:

hyperlink

All Results (no minimum matches)

Paste in a query sequence to find its location in the genome. Multiple sequences may be searched if separated by lines starting with '>' followed by the sequence name.

Human (hg38) BLAT Results

BLAT Search Results

Custom track name:

Custom track description:

ACTIONS	QUERY	SCORE	START	END	QSIZE	IDENTITY	CHROM	STRAND	START	END	SPAN
<a href="#">browser details</a>	YourSeq	146	1	146	146	100.0%	chr2	-	25229049	25229194	146
<a href="#">browser details</a>	YourSeq	21	81	102	146	100.0%	chr12	-	18672011	18672033	23

# Case #3

- Checked the kit - 3' vs 5' (matched)
- Checked the data – blatted a read at random (matched to human)
- Checked 10x indices – do they appear in the whitelist (yes)
- Pulled our hair out, contacted production

# Case #3

- Retraced our steps double checking all of our work
- Blatted a few more reads
  - They all matched poorly to the human genome
  - They all matched well to the mouse genome

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**dx: SPECIES MIXUP**

- Just by chance, the first read we checked was from a very highly conserved gene!

# Damage Control

- Realign to the correct species
  - Gave expected high alignment rate
- Still have to resolve what happened with the sample naming
  - Was the species designation just wrong?
  - Is the entire sample named wrong? (swap)

# Xenograft contamination

- Related topic is dealing with Xenograft data
  - e.g. human tumors implanted in a mouse
- Mouse reads with homology to human genome
- One solution is Xenosplit – alignment-based read filtering
  - Human, mouse, ambiguous

# Global alignment/mismatch issues

- Sample swaps
  - check SNP concordance
  - Somalier is a tool for rapid sample identity checking
- Contamination
  - VAFs, IGV inspection are your friends
- Species swaps
  - Check a few reads, some concordance is expected!

# Case #4

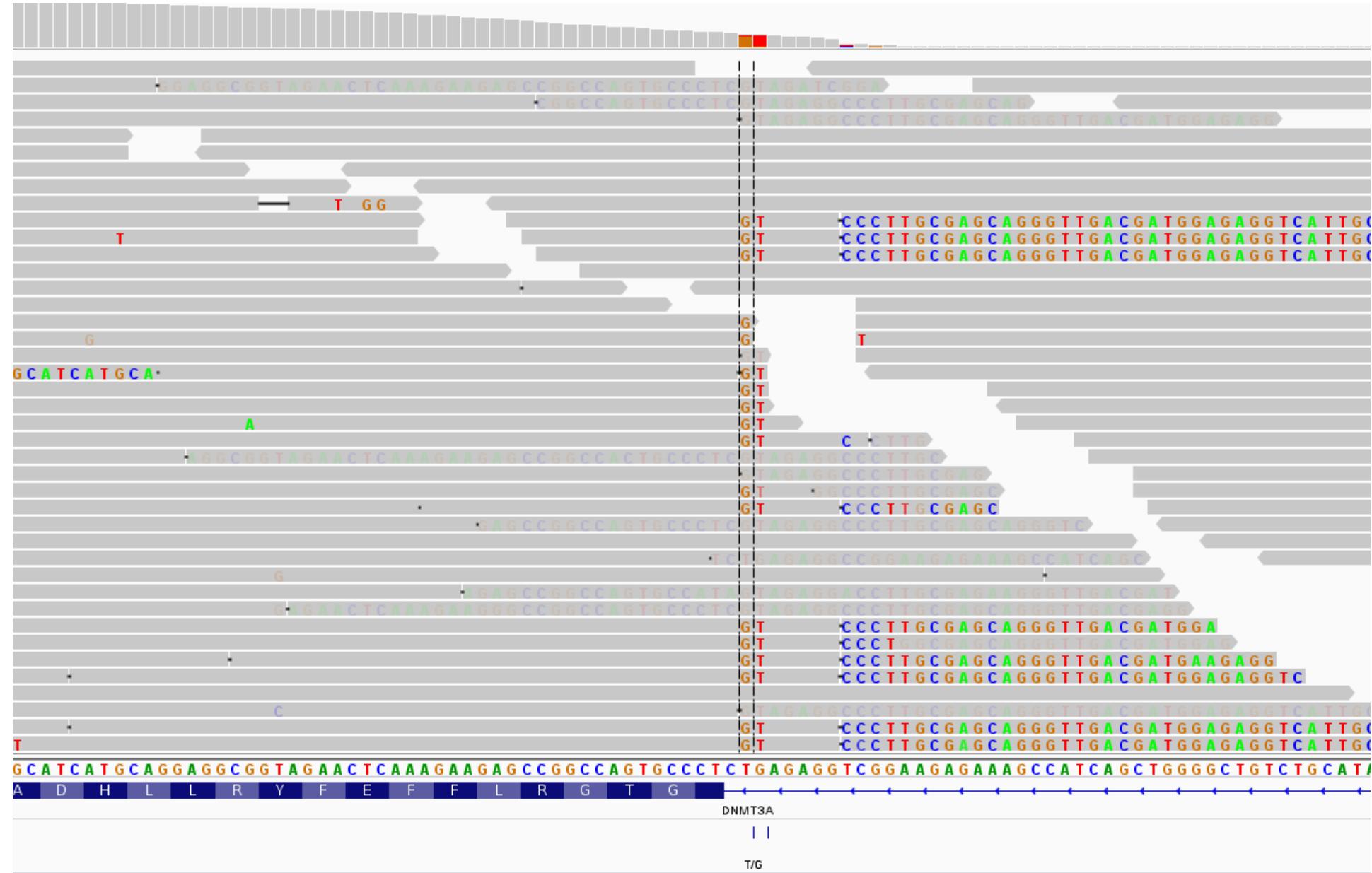
Exome sequencing of a tumor sample

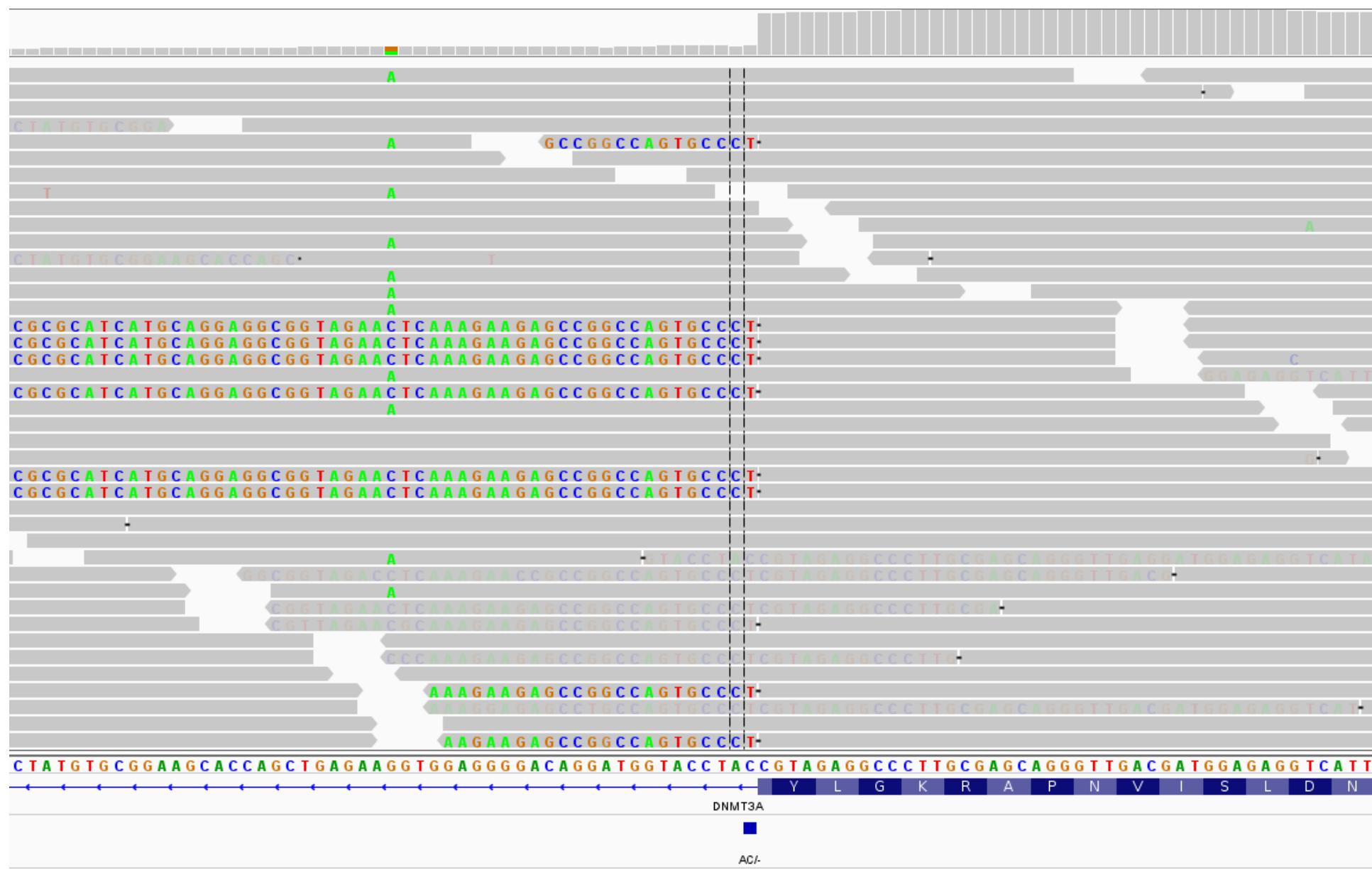
Mutations sorted by the number of times a single gene is hit

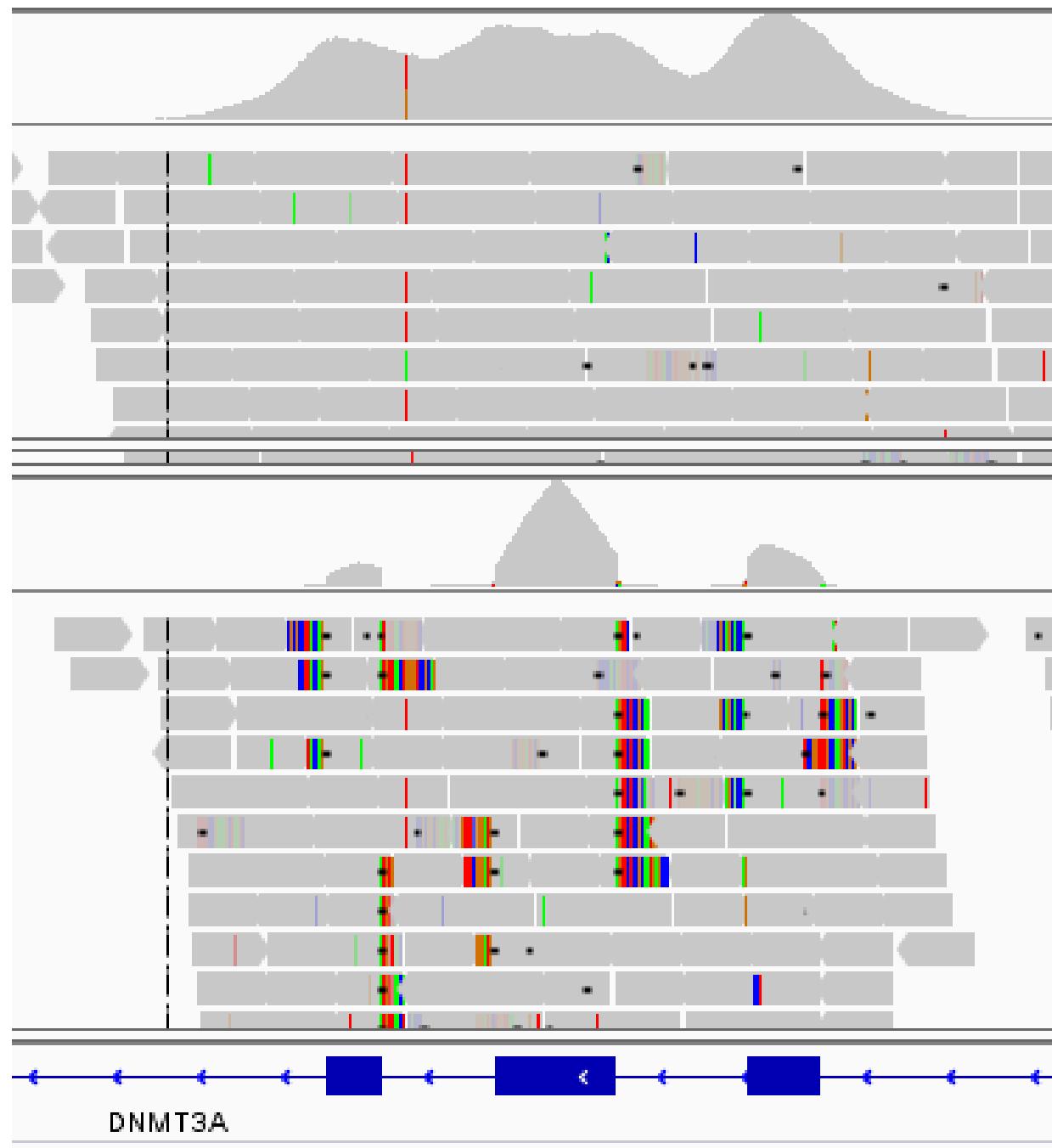
43	DNMT3A
2	WT1
2	SLC35F3
1	UNC93B5
1	TSLP
1	TRPS1
1	TARDBP
1	SUN3
1	SREBF1
1	SPTBN2
1	SPAST
1	SNX1
1	SNRNP40
1	SLC17A3
1	SELK
1	RUNX1
1	RCC1
1	PTPN11

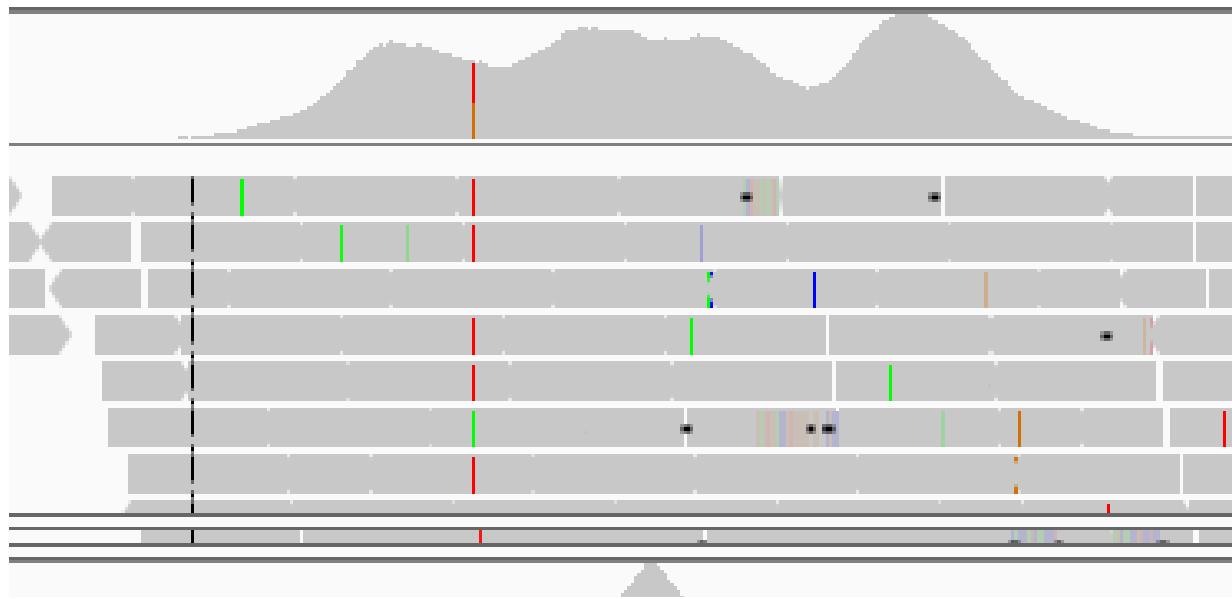
# Variants called in DNMT3A

2	25457242	C	T	DNMT3A	missense	p.R882H
2	25458572	G	A	DNMT3A	splice_region	e21+4
2	25463169	A	G	DNMT3A	splice_site	e18+2
2	25463170	C	A	DNMT3A	splice_site	e18+1
2	25463321	T	G	DNMT3A	splice_site	e18-2
2	25463322	G	T	DNMT3A	splice_region	e18-3
2	25463507	AC	-	DNMT3A	splice_site_del	e17+1
2	25463600	C	A	DNMT3A	splice_site	e17-1
2	25464428	T	G	DNMT3A	splice_region	e16+3
2	25464430	C	T	DNMT3A	splice_site	e16+1
2	25467021	C	G	DNMT3A	splice_region	e14+3
2	25467022	A	T	DNMT3A	splice_site	e14+2
2	25467211	G	C	DNMT3A	splice_region	e14-4
2	25467213	A	G	DNMT3A	splice_region	e14-6
2	25467403	CCT	-	DNMT3A	splice_region_del	e13+4
2	25467524	A	T	DNMT3A	splice_region	e13-3
2	25467526	A	C	DNMT3A	splice_region	e13-5
2	25469028	C	T	DNMT3A	splice_site	e10+1
2	25469181	G	T	DNMT3A	splice_region	e10-3

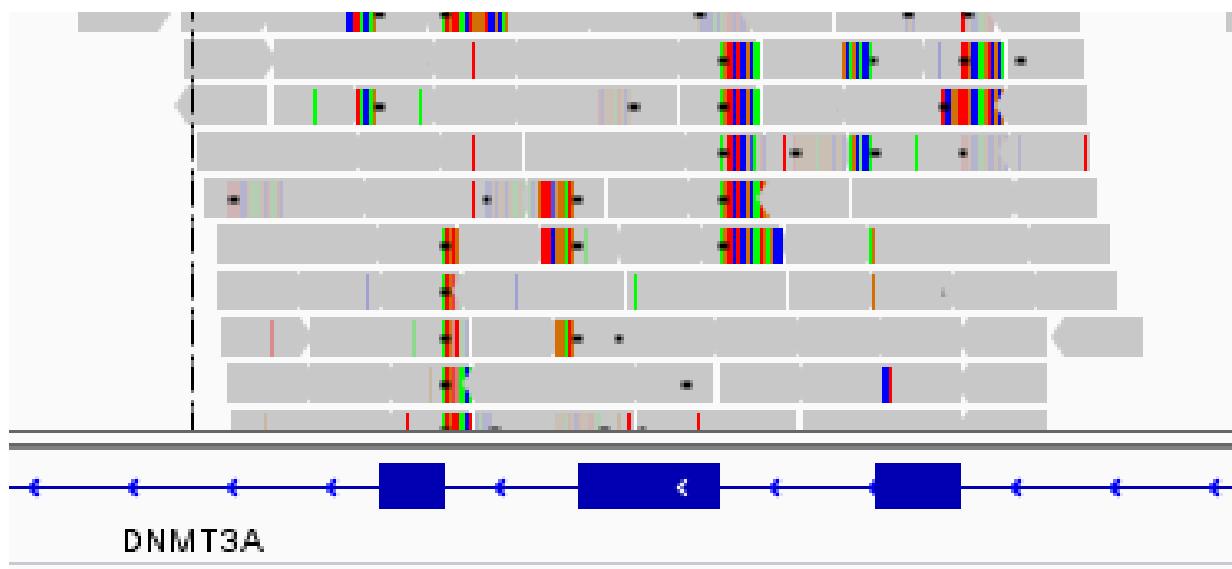








# dx: cDNA contamination



# Damage Control

- If it is limited to a single gene (as in this case), could remove all splice-site adjacent mutations in that gene
  - If it's many genes/widespread, you might miss a lot of real events!
- Remake the libraries, resequence the sample

# Case #5 – WGS serial samples

```
$ wc -l relapse_mutations.txt
```

```
10114 relapse_mutations.txt
```

# Case #5 – WGS serial samples

```
$ wc -l relapse_mutations.txt
```

```
10114 relapse_mutations.txt
```

Max from TCGA AML cohort:

1298

# Case #5 – WGS serial samples

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10114 relapse_mutations.txt
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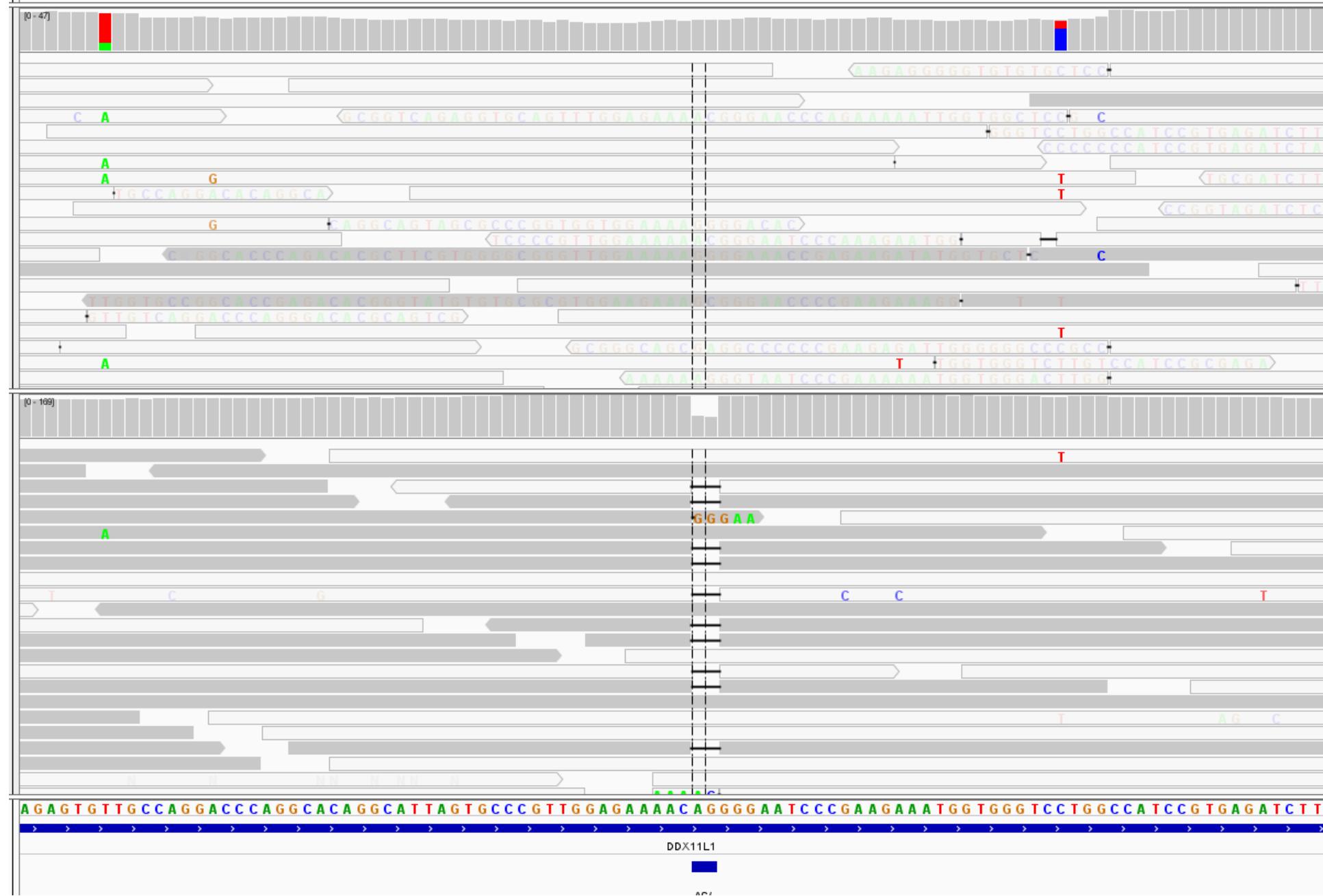
Max from TCGA AML cohort:  
1298

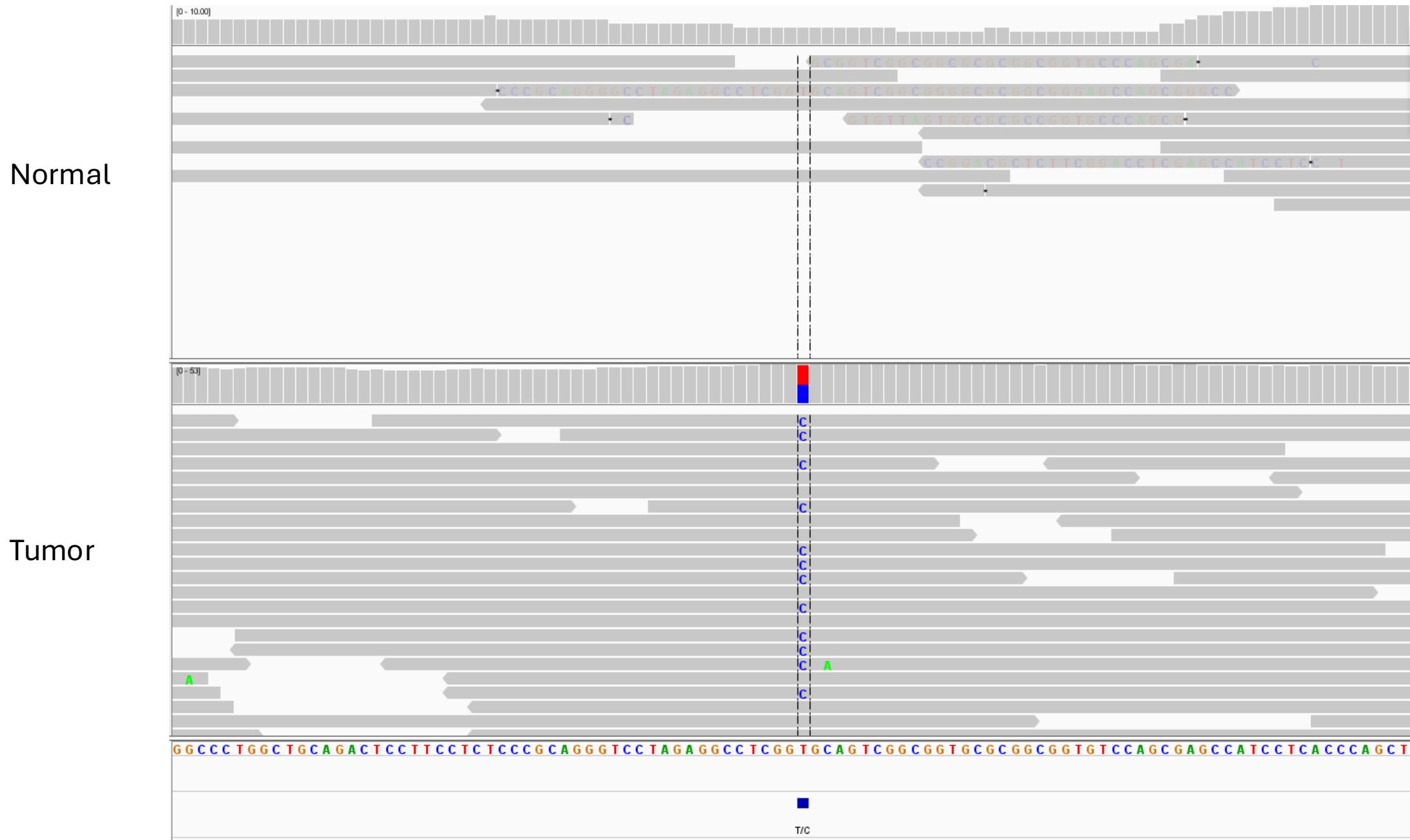
Primary tumor from this sample: 573

# Case #5 – WGS serial samples

- Sample swap?
- Mismatch repair?
- Mutagenic therapy?

Normal

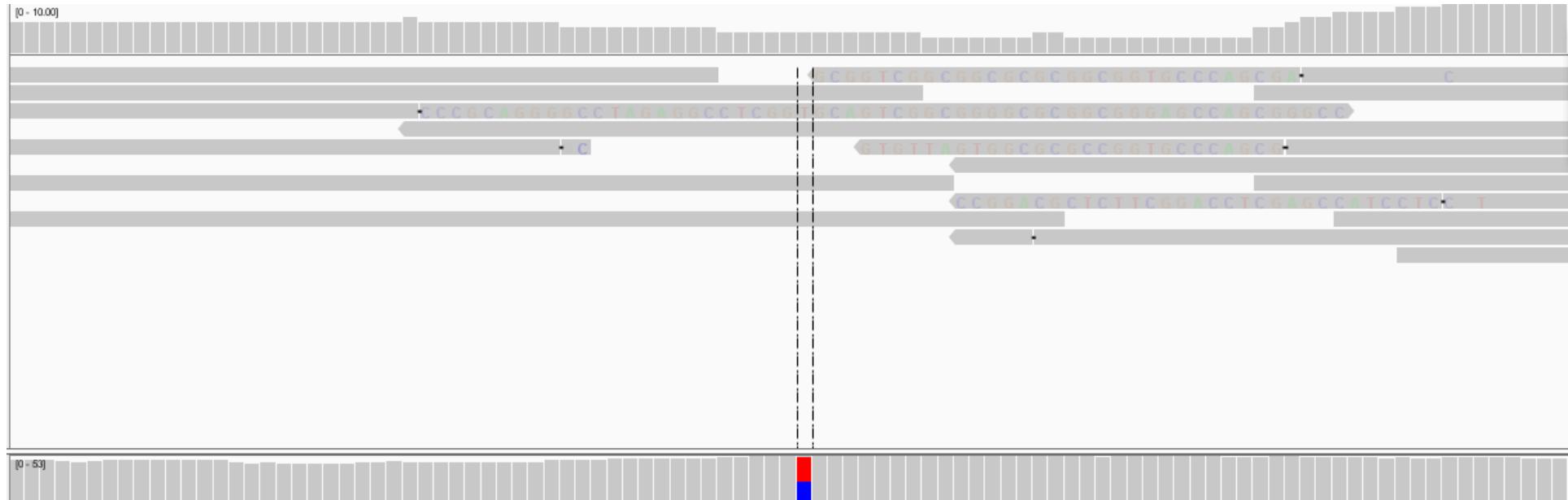




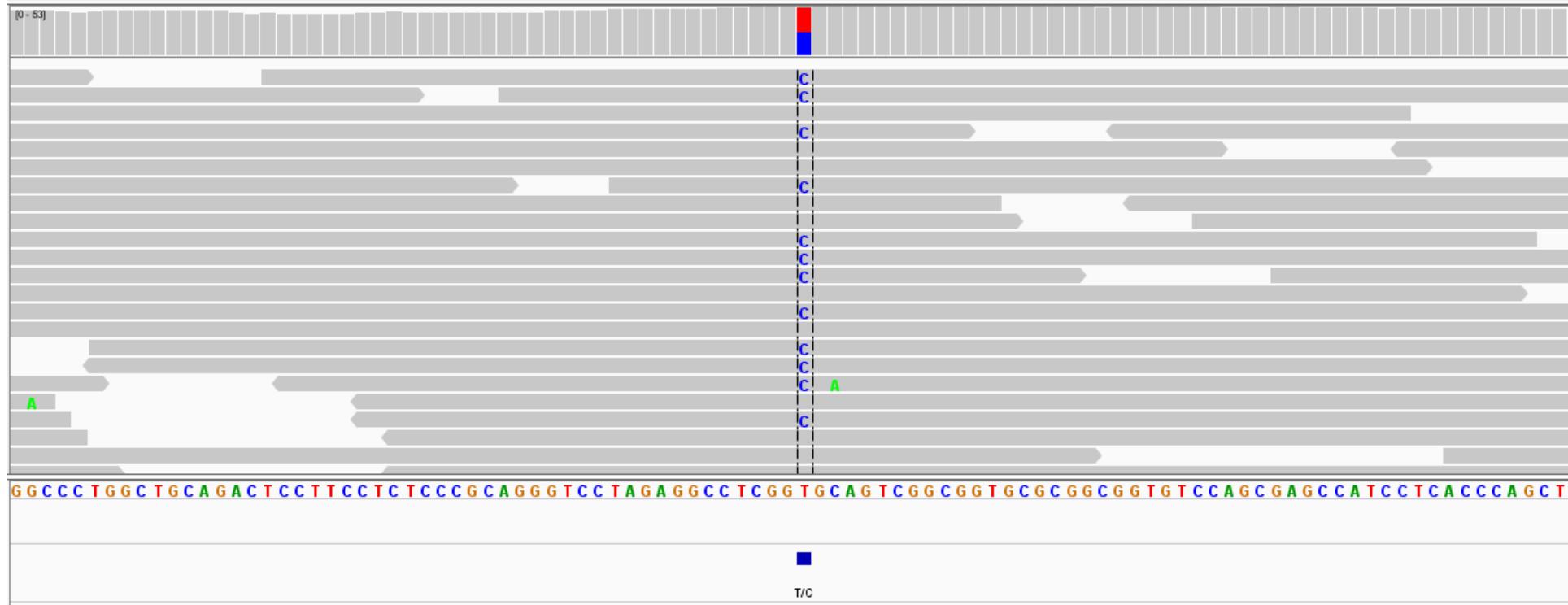
- Normal – sequenced with a mix of 75bp and 100bp PE reads
- Relapse – sequenced 6 years later with 125bp PE reads

**dx: poorly matched controls**

Normal



Tumor



# Damage Control

- Resequence the normal with matching read lengths
- Match your data as closely as possible!
  - Read lengths
  - Capture kits
  - Sample preparation

# General Tips

- Visualize your data
  - A picture is worth a thousand p-values
- Hone, and then trust your instincts
  - If something seems unusual, it's often either a big problem or a big finding
- Be relentless
  - don't stop digging until you convince yourself that nothing is wrong

“

An expert is a man who has made all  
the mistakes which can be made, in a  
narrow field.

--Niels Bohr