



**BC Cancer Agency**

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# **Bioinformatics Series: Designing An NGS Study For My Biological Question**

Yussanne Ma

Genome Sciences Centre

# Genome Sciences Centre

## Sequencing

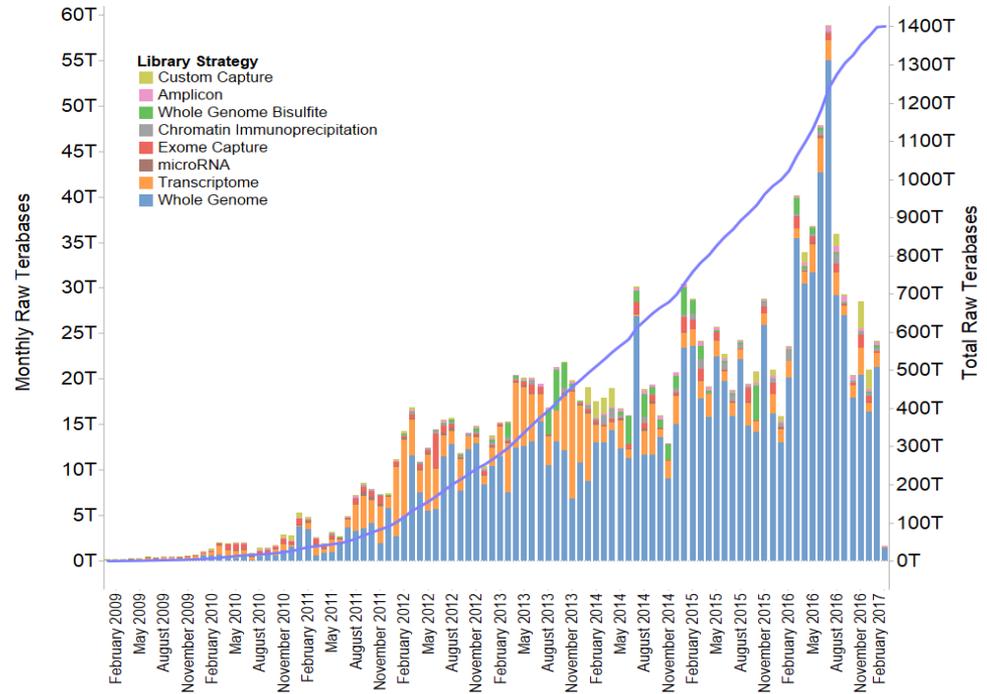
- 5 Illumina HiSeqX
- 4 Illumina HiSeq2500
- 2 NextSeq500
- 3 Illumina MiSeq
- 1 Life 3730 xl

1500 libraries per month  
>80Tbases per month



## Compute

- 2 secured data centres
- Compute clusters : 800 nodes, 24,000 hyper-threaded cores
- 48 - 384 GB RAM per node
- High memory (1.5TB RAM) computers
- >11 Petabytes on-line disk storage



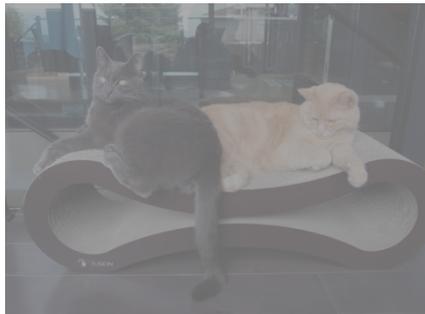
Engaged in over 50 ongoing projects and collaborations from experimental design to data interpretation



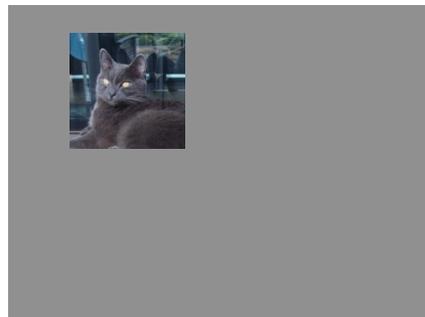
# Why experimental design is important



Analysis and interpretation of sequencing data is completely dependent on everything upstream



Sample quality, sequencing and type of sequencing matter



The area being sampled matters

Bioinformatic corrections can be made but it's always best to plan ahead



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# Outline

Genome sequencing

Transcriptome sequencing

Integrative approaches

Other technologies



# Outline

## Genome sequencing

Genotyping arrays

Exome and custom capture

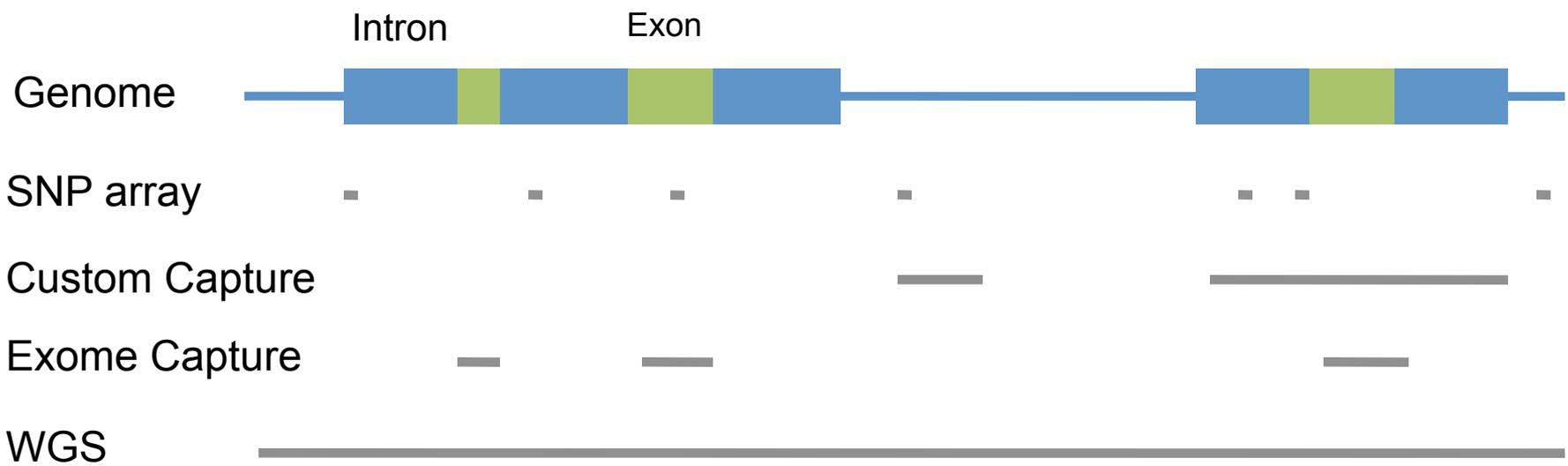
Amplicon

Whole genome sequencing

Population size and controls

Factors affecting quality of variant calls

# Genome sequencing overview

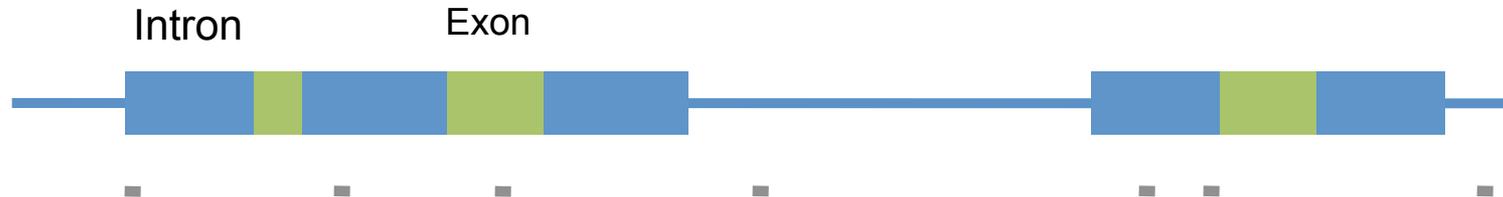


There are many ways to subsample the genome  
 The cost trade-off is between area covered and depth

Sometimes the genome can be overkill



# Genotyping arrays



Sampling of the genome at locations of known single nucleotide polymorphisms using intensity probes

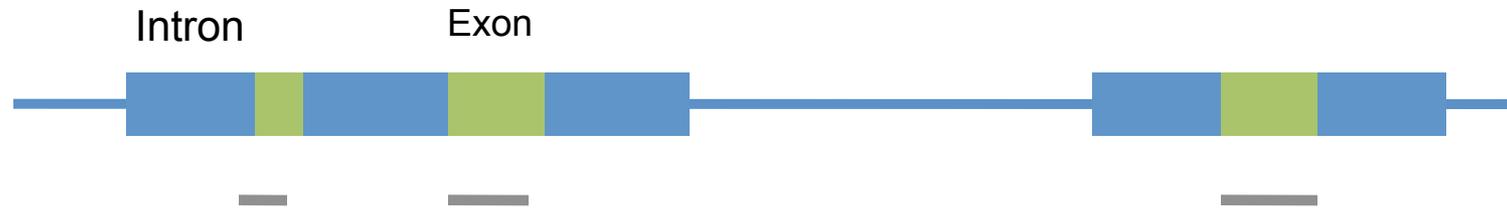
Used for: Studying common variants in large number of cases and controls

Limitations: Cannot be used for calling of rare or novel variants, or structural variants. Resolutions for copy number variation are low

Example project: Genome-wide association study to look for inherited cancer susceptibility loci



# Exome and custom capture



Probes are used to capture all exons, or a specific set of genomic regions

Used for: Studying only coding changes, or a those found in the pre-defined area of interest

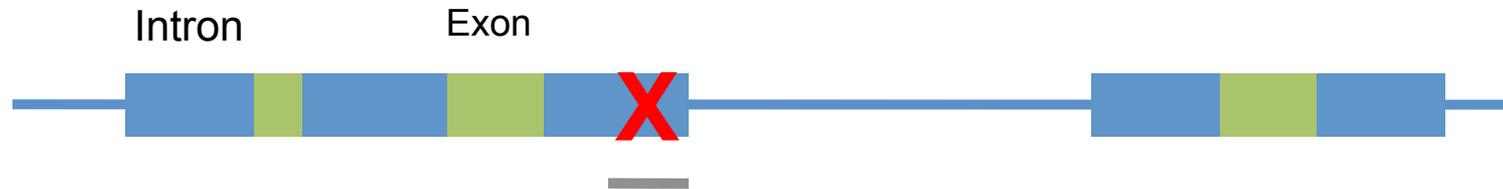
Limitations: Cannot call variants outside of capture area.

Copy number and structural variants are difficult to call

Example project: Discovery of recurrently mutated genes in large cohort, clinical panel



# Amplicon and Sanger sequencing



Primers are designed that span a genomic event, or sequence across an event.

Used for: Determining the presence or absence of specific events (SNVs, indels, SVs)

Limitations: Cannot be used for discovery, need exact breakpoints in most cases

Example project: Orthogonal verification of putative event discovered by WGS to benchmark tools, determining presence of metastatic fusion event in primary sample.



# Whole genome sequencing

Used for: Full characterization of genome including

Novel genes and events

- SNVs and indels not seen in the population

- Private events in recurrent genes or pathways

Complex events

- Copy number

- Structural variants

Genomic landscape of a population

- Mutation signatures

Limitations: Sample size and depth due to cost

Best used for studies with no *a priori* knowledge of population samples, in depth study of single patient tumour



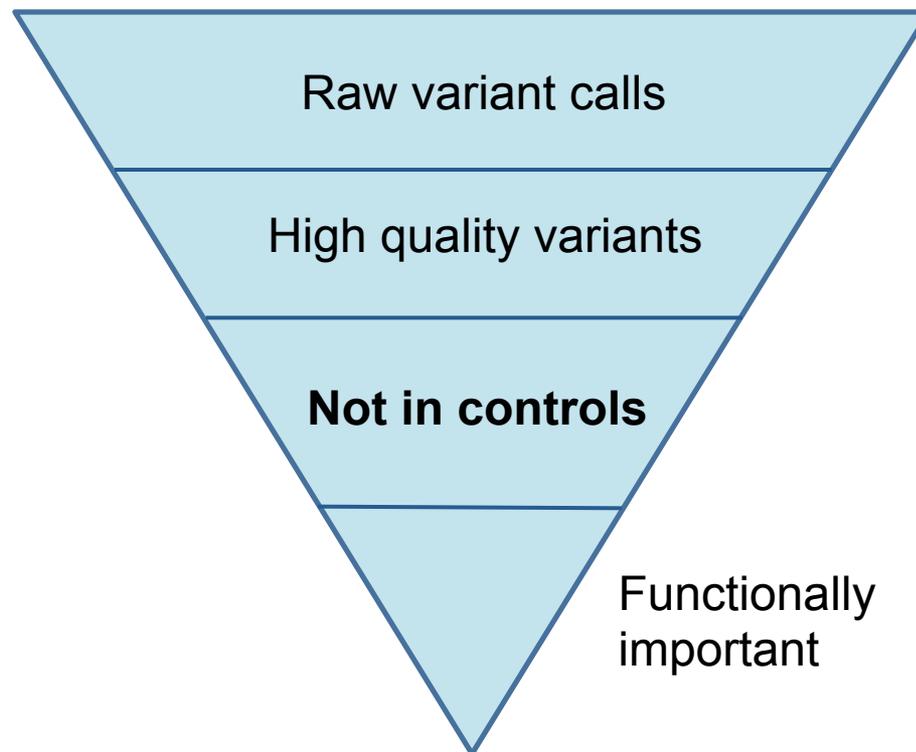
# Genome sequencing: summary

| Technology       | SNVs                 | CNVs             | SVs                  | Mutational burden | Mutational landscape |
|------------------|----------------------|------------------|----------------------|-------------------|----------------------|
| WGS              | +++                  | +++              | +++                  | +++               | +++                  |
| Exome            | +++<br>(coding only) | +<br>(coding)    | +<br>(coding)        | ++                | -                    |
| Custom capture   | +++<br>(on target)   | +<br>(on target) | +<br>(on target)     | +                 | -                    |
| Genotyping array | Specific events only | +                | -                    | -                 | -                    |
| Amplicon         | Specific events only | -                | Specific events only | -                 | -                    |
| Sanger           | Specific events only | -                | Specific events only | -                 | -                    |



# Population size and controls

3 million germline variants, 10,000-100,000 somatic variants on average per sample





# Population size and controls

GWAS: Large sample size needed to achieve statistical significance, 1:1 cases and controls

Rare disease: Sequencing of parents reveals patterns of inheritance, sequencing of unaffected relatives helps to filter out passengers

Somatic variants: Matched normal is needed to filter out passenger mutations

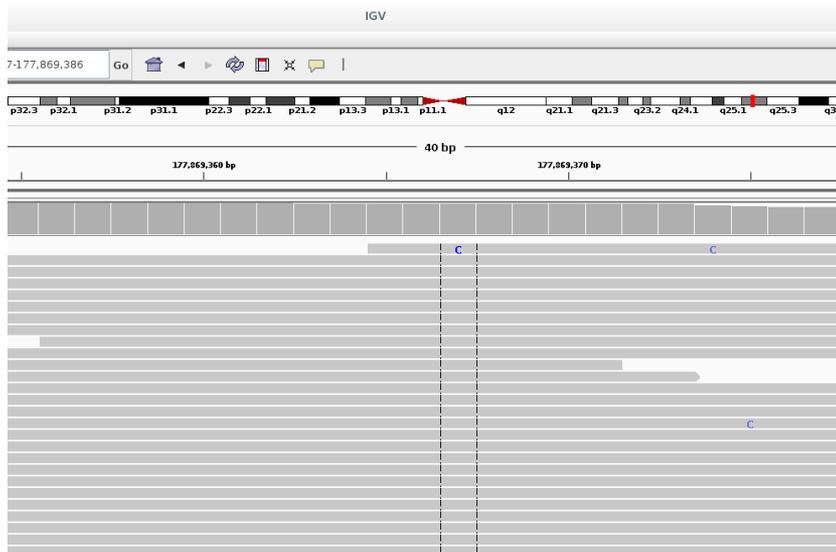


# Factors affecting quality of variant calls: sequencing depth

Why 30X genome?

Rule of thumb: it takes 3-10 high quality reads to call a variant

Need to account for variable coverage, evenness of coverage, tumour content, ploidy





## Factors affecting quality of variant calls: sequencing depth

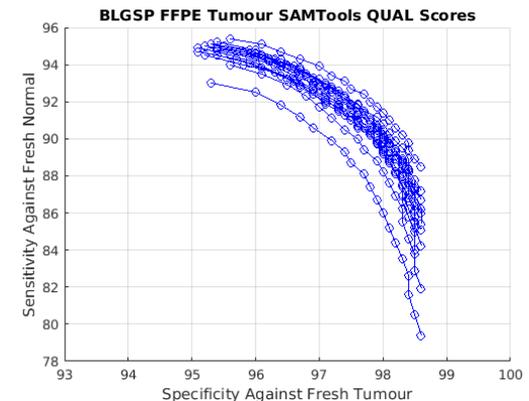
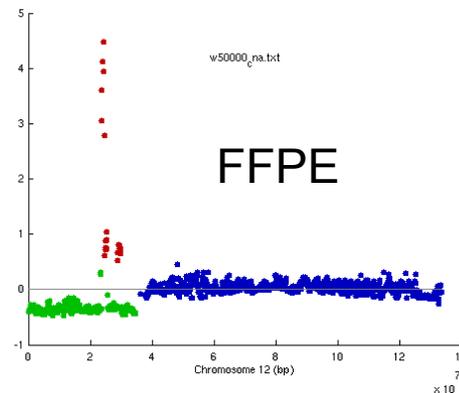
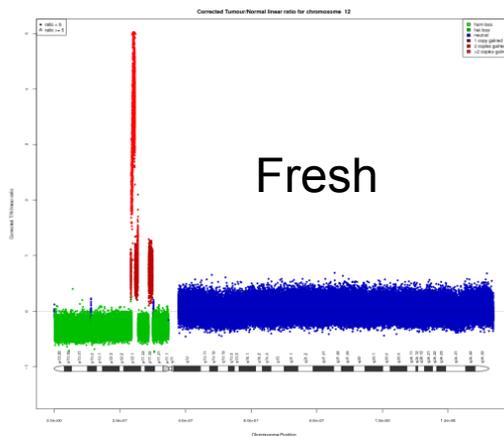
| Type of variant                      | Depth needed |
|--------------------------------------|--------------|
| Germline, diploid                    | 30X          |
| Tumour > 70% tumour content          | 30-40X       |
| Tumour 40 - 70%                      | 40-60X       |
| Low tumour content, subclonal events | > 100X       |



# Factors affecting quality of variant calls: sample type

## FFPE vs Fresh frozen

All of our protocols (WGS, RNA, miRNA) can be run on FFPE samples, but they may result in slightly lower yield and diversity, and a higher false positive rate for SNV and SV detection, as well as noisier CNV calls





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# Outline

## Transcriptome sequencing

RNA sequencing

miRNA sequencing

Batch effects



# RNA sequencing

## Ribosomal depletion vs. polyA selection

no ribosomal RNA  
captured

non-polyadenylated  
transcripts are captured

lower minimum input  
requirement

higher intergenic and  
intronic content

higher ribosomal RNA  
content

only polyadenylated  
transcripts are captured

higher minimum input  
requirement

lower intergenic and  
intronic content



# RNA sequencing

Used for:

Gene, exon and isoform-level quantification

Quantifying expression of genomic events (SNVs, SVs)

Detecting novel transcripts

Detecting RNA edits

Differential expression between groups (condition/tissue/  
tumour type) to identify expression markers

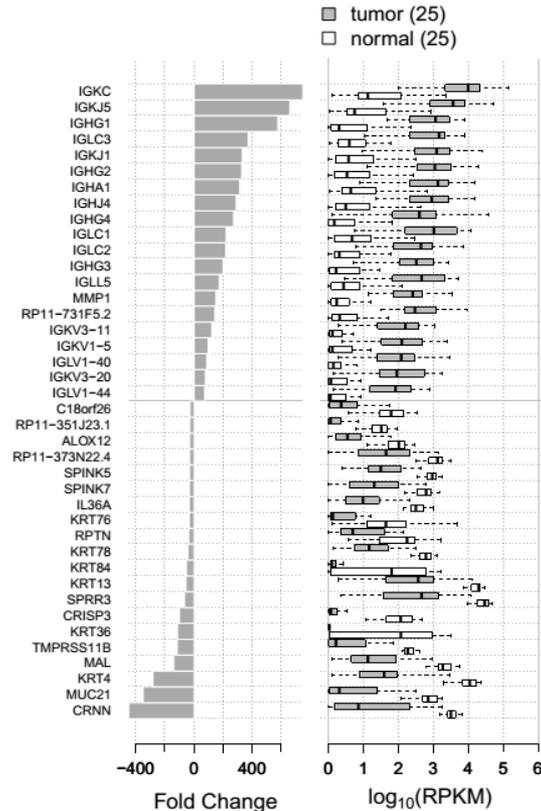
Correlation and clustering of samples by gene expression to  
identify subgroups



# RNA sequencing

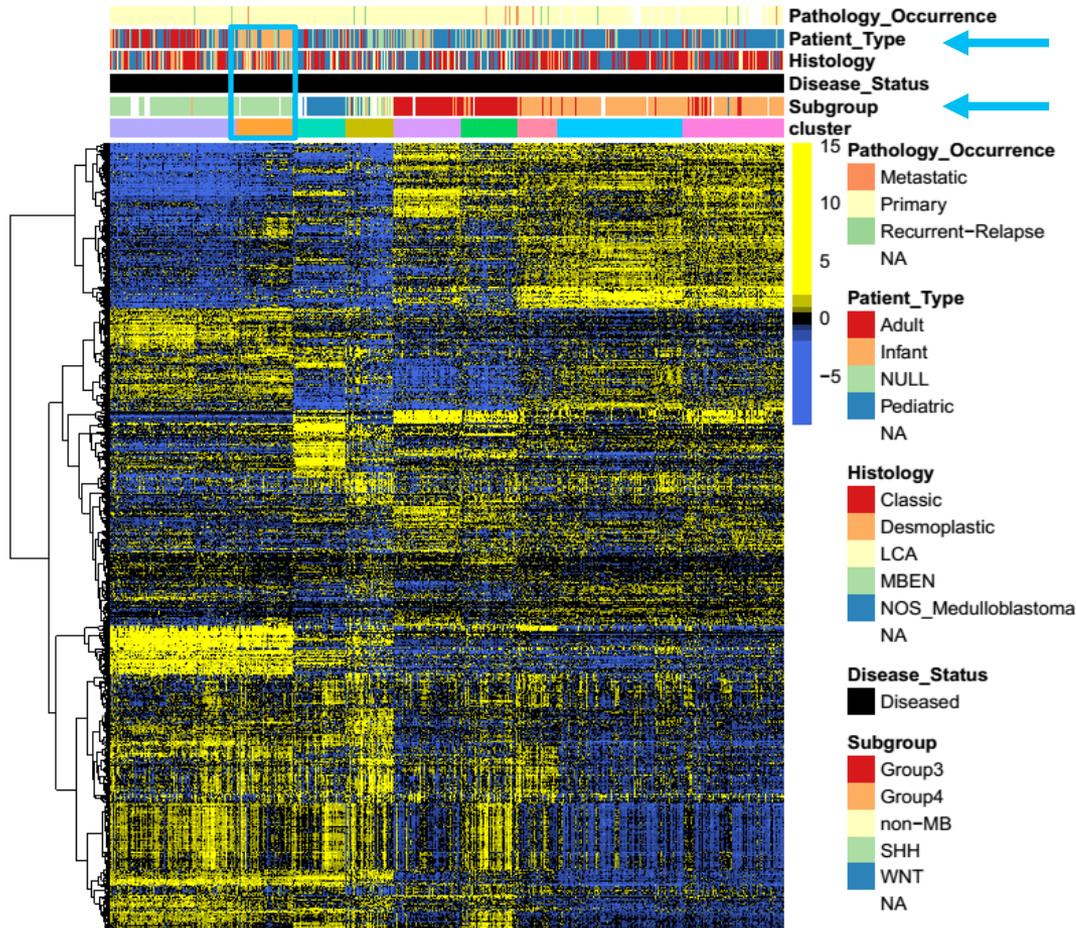
Differential expression between tumour groups

Results are more difficult to interpret with low sample size





# RNA sequencing



Hierarchical clustering of medulloblastoma samples by gene expression

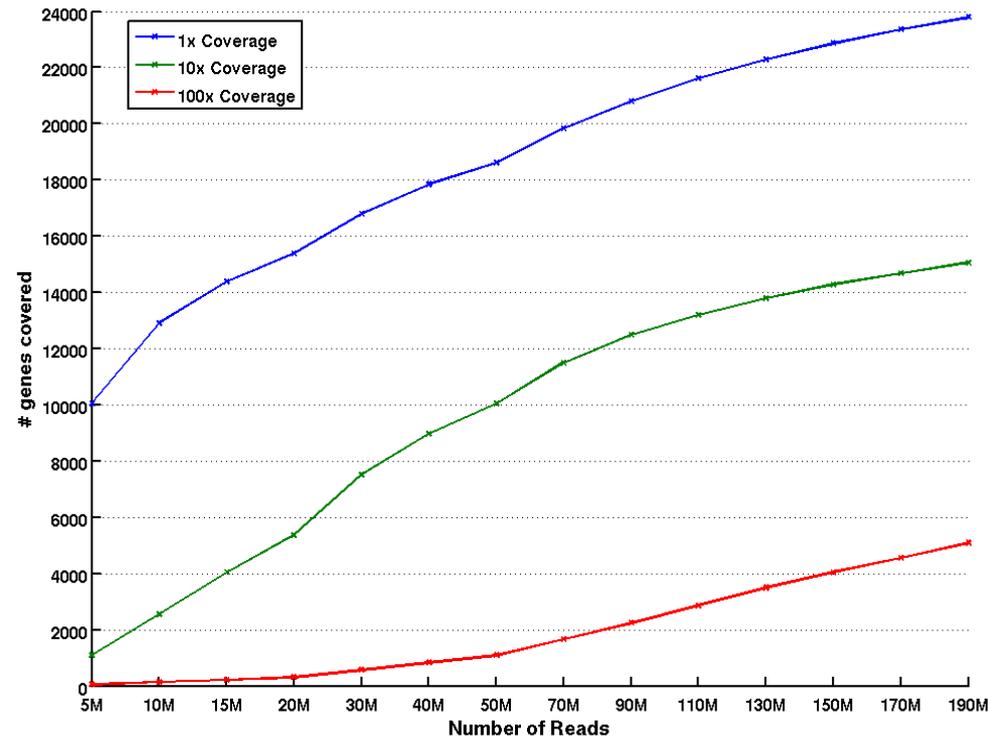
Samples cluster by subgroup

Within a subgroup samples cluster by patient type

Most informative with large sample size and detailed covariates eg. clinical data



# Sequencing depth



Gene diversity for UHR control at different levels of downsampling

Diversity does not tend to saturate



# Sequencing depth

| Number of reads per library                                                  | 200M   | 120M   | 60M                  | 40M    |
|------------------------------------------------------------------------------|--------|--------|----------------------|--------|
| Number of genes at 1X                                                        | 23,000 | 20,000 | 18,000               | 15,000 |
| Number of genes at 10X                                                       | 14,000 | 12,000 | 10,000               | 5,000  |
| Expression quantification                                                    | ++     | ++     | ++                   | ++     |
| Differential expression                                                      | ++     | ++     | ++                   | ++     |
| Known transcript quantification                                              | ++     | ++     | ++                   | ++     |
| Detection of structural variants with gene partners or breakpoints specified | ++     | ++     | ++                   | +      |
| Detection of SNVs and small indels with known coordinates                    | ++     | ++     | ++                   | ++     |
| De novo SNV calling                                                          | ++     | ++     | +                    | -      |
| De novo structural variant calling                                           | ++     | +      | Alignment based only | -      |



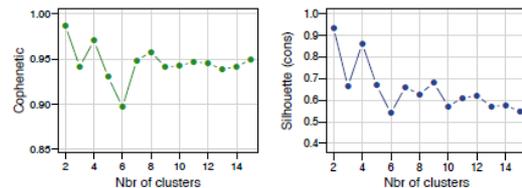
# miRNA sequencing

Used for:

Quantification of miRNA expression

Differential expression and expression clustering

Correlation with gene expression to identify targets

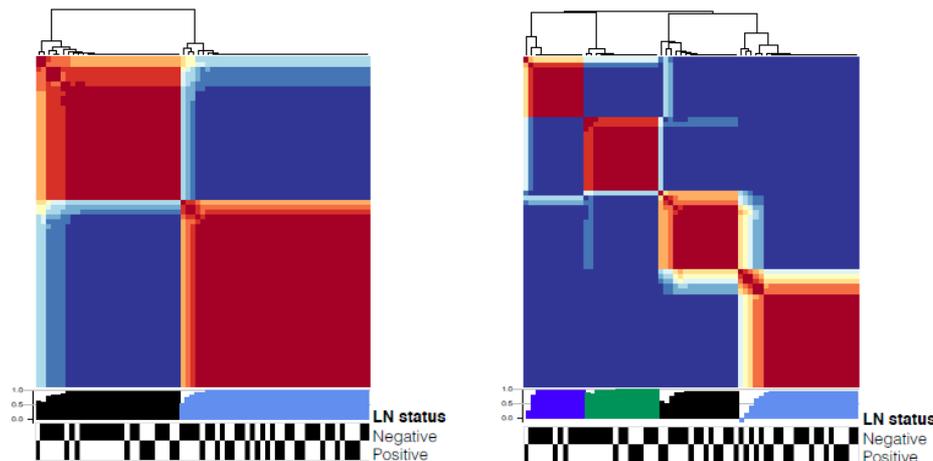


**Using miRNA to determine a signature for prognosis**

miRNA clustering is found to be more sensitive to subgroups

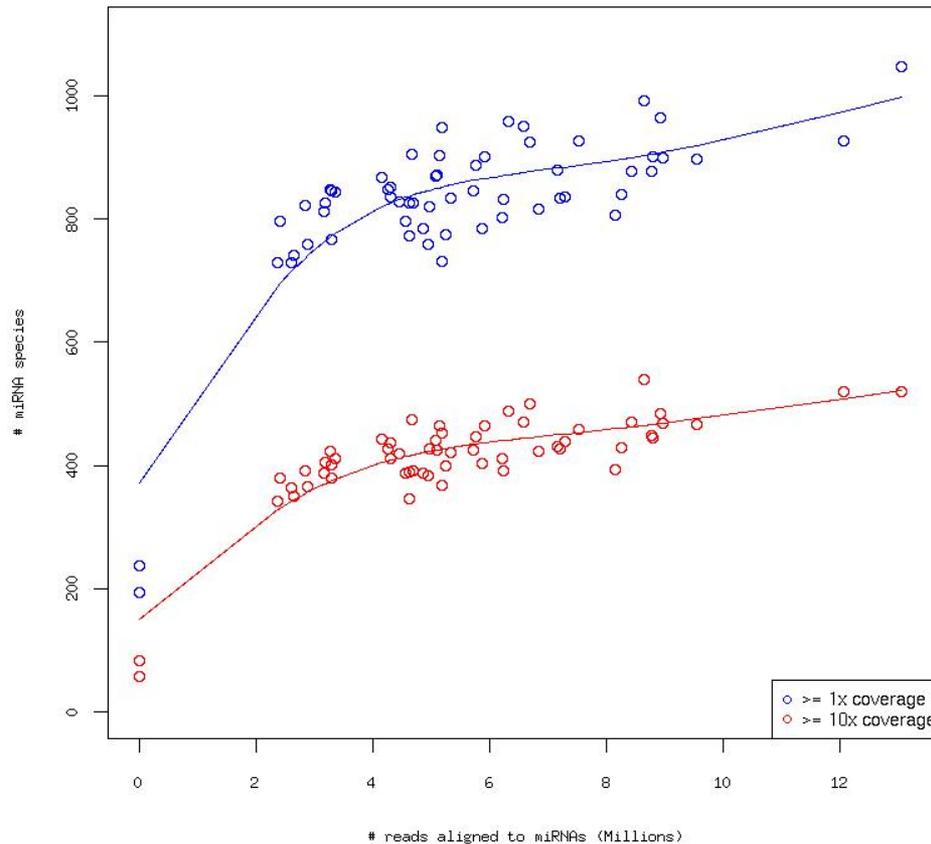
Search space is smaller and cost of sequencing is lower

May be easier to translate into clinical test





# Sequencing depth



miRNA diversity vs  
number of reads  
aligned to miRNA

Two failed samples on  
far left

Saturation between 2  
and 4 million reads

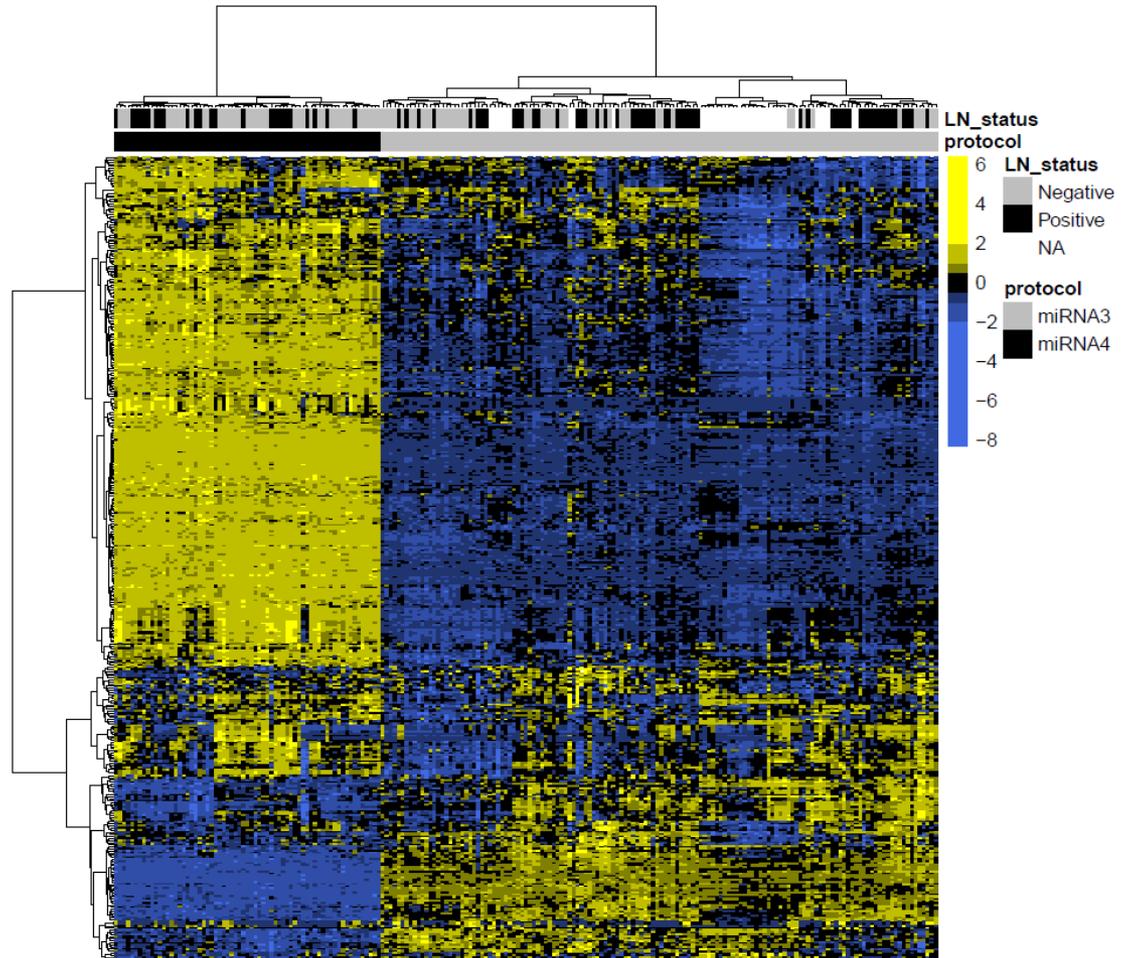


# Batch effects

Samples cluster by protocol, so clustering is difficult to do across multiple protocols

Sample sets sequenced using different protocols are best used as validation, or for meta analysis

Batch effect correction is the most effective with technical replicates





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# Outline

## Integrative approaches

Genome and transcriptome sequencing

Clonal evolution experiment

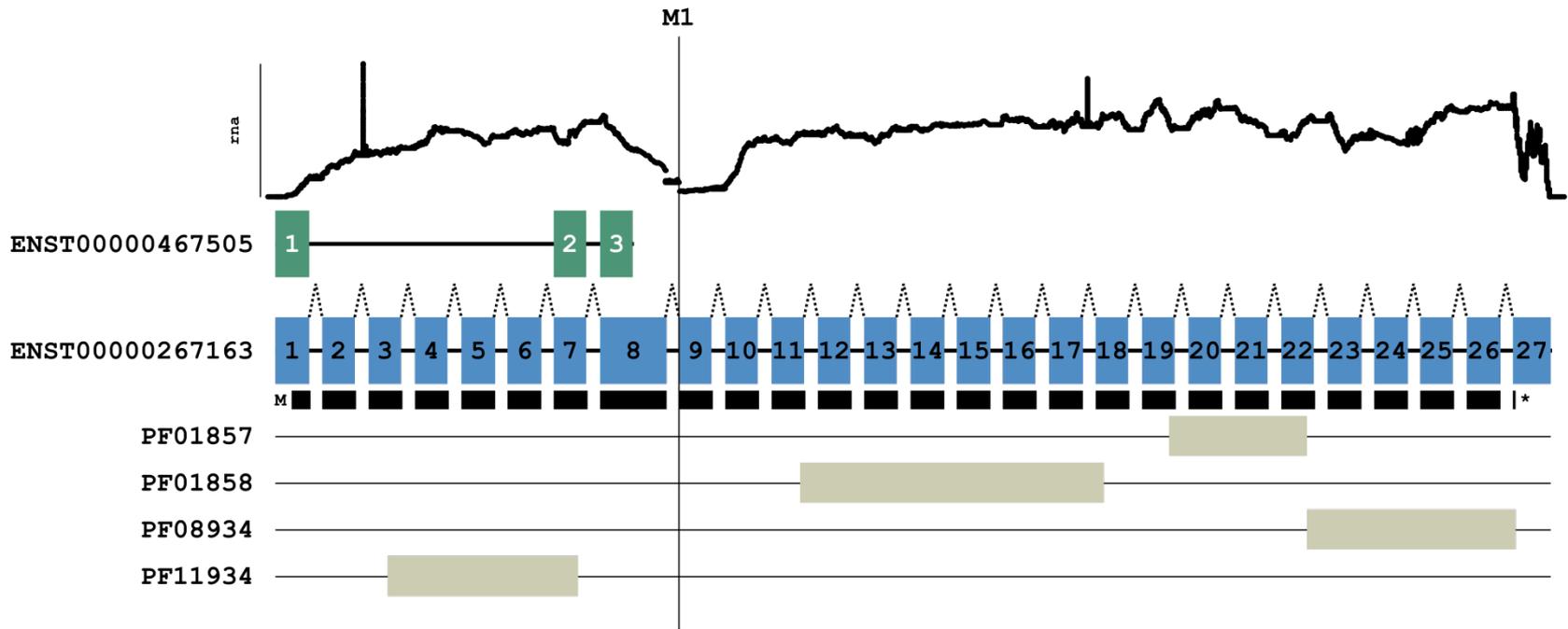
Integrative analysis to study 'dark matter' in cancer



# Integrative approaches: genome and transcriptome

RNAseq provides orthogonal validation of genomic events

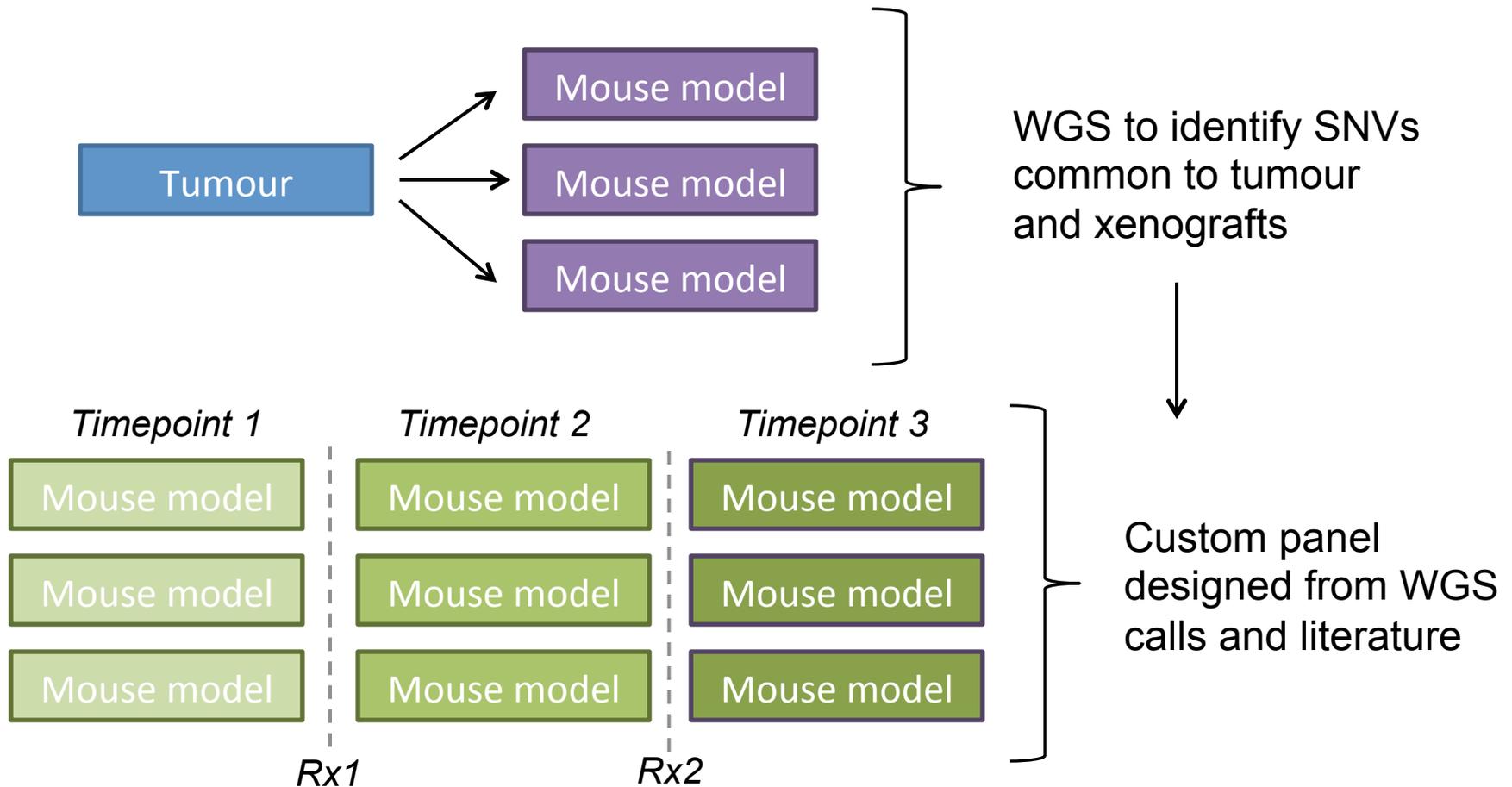
Combined approach improves specificity, and can identify/confirm alternative splicing and elucidate the effects of genomic events on transcription



Alternative splicing at M1 is identified in the structural variant analysis of RNA and DNA, and gene expression data confirms exon 9 skipping event



# Integrative approaches: Clonal evolution

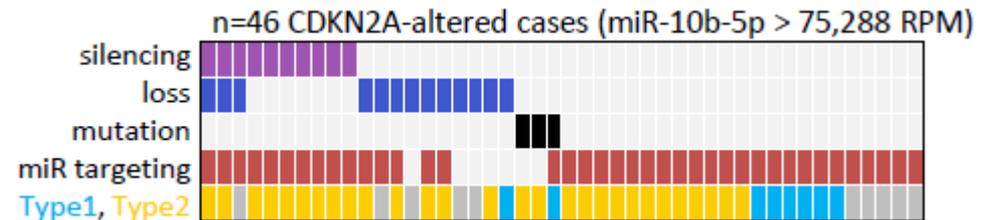
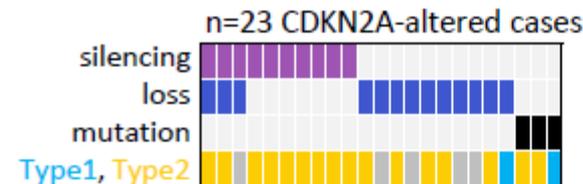


Clonal evolution over multiple timepoints and treatment events



# Integrative approaches: 'dark matter'

miRNA, RNA and WGS and methylation sequencing identify multiple mechanisms in which CDKN2A function is disrupted in papillary renal-cell carcinoma





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# Outline

## Other technologies

Microbial analysis

de novo genome assembly

Single-cell sequencing

Epigenomics

Immunogenomics

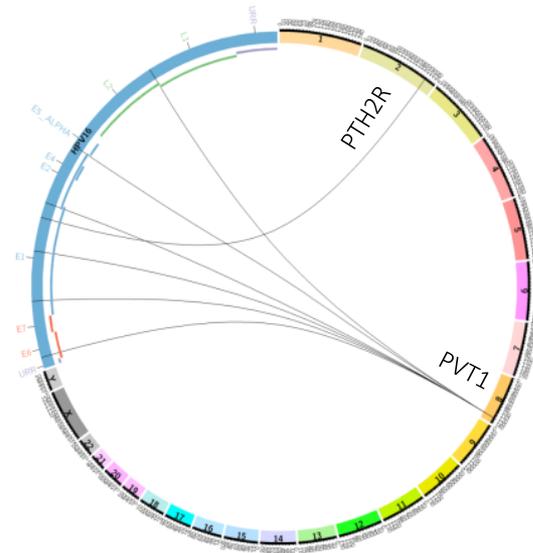


# Microbial analysis

16S sequencing: Identification and quantification of known bacterial species. Useful for survey of large number of samples

Short read sequencing (shallow): Rapid classification of known microbial species in metagenomic samples

Whole genome and transcriptome sequencing: Microbial expression and genome integration in tumour samples





# De novo genome assembly

Short read sequencing at ~30X is sufficient for de novo assembly using ABySS to produce contigs

Contigs can be:

- Aligned to existing references to identify variants in new strains

- Annotated to identify putative genes

Extension to a full draft reference will require additional sequencing to build scaffolds:

**Mate-pairs:** large insert, long reads to extend assembly

**10X Chromium:** Phased genomes, localized assemblies

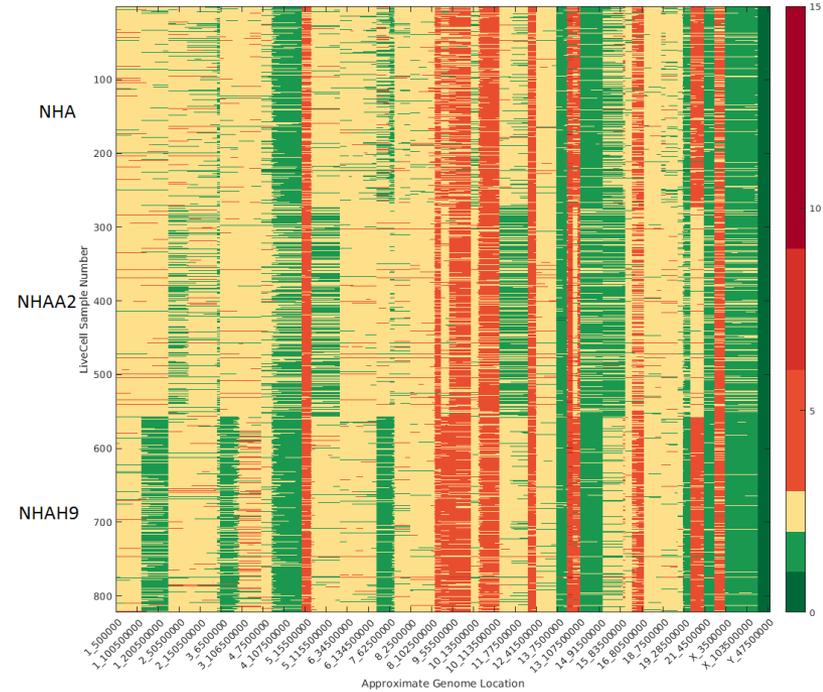
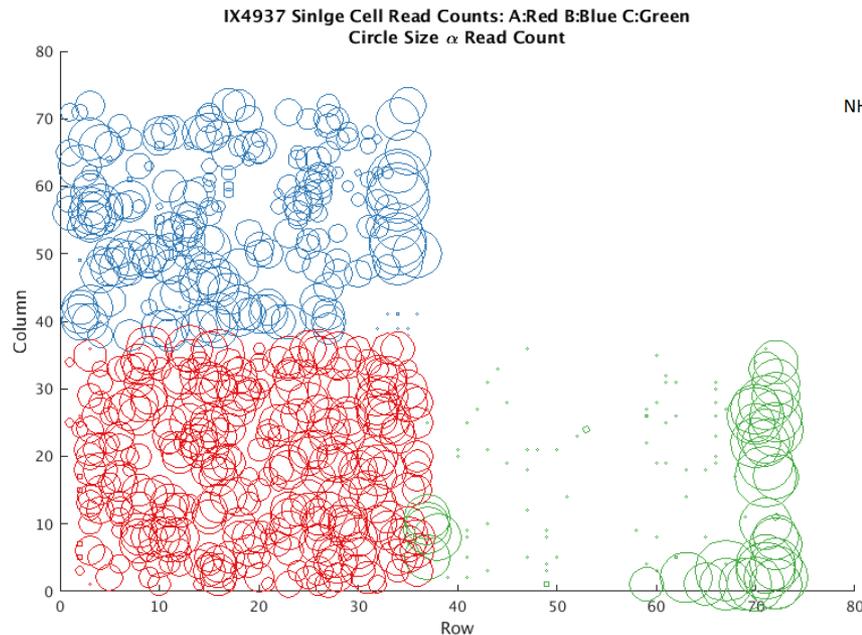
**Oxford nanopore:** high throughput long reads

**Pacbio:** consensus long read with lower error rate



# Single-cell sequencing

WGS and RNA sequencing from individual cells allows for single cell resolution of copy number and expression



Cell populations treated under different conditions can be examined separately



# Epigenomics

Post-transcriptional modification cannot be detected through genome and transcriptome sequencing

Efforts such as the International Human Epigenome Consortium have provided comprehensive datasets for comparison and interpretation of epigenomic data

ChIPseq and bisulphite sequencing (array, whole genome or capture) are used to study histone modification and DNA methylation

Examples of analysis: Identify genes and pathways that are epigenetically modified, correlated ChIP data with expression and mutational data, cluster samples by DNA methylation profile



# Immunogenomics

TCR/BCR sequencing

HLA typing

Analysis from WGS and WTS sequencing:

T and B cell repertoire

HLA typing

Cell type abundance

Neoantigen prediction



# How much disk do I need

| Data*                    | Typical file size |
|--------------------------|-------------------|
| 30X genome               | 50G               |
| Full-depth transcriptome | 15G               |
| miRNA                    | 500M              |
| 1 lane Hiseq 2500        | 50G               |
| 1 lane Hiseq X           | 65G               |
| Variant files            | 10-100M           |

\*human data, bam/fastq.gz/raw assembly data are similar in size



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# Questions

About this talk: [yma@bcgsc.ca](mailto:yma@bcgsc.ca)

About sequencing and bioinformatics at the  
GSC: [dmiller@bcgsc.ca](mailto:dmiller@bcgsc.ca)



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