

DNA methylation data analysis

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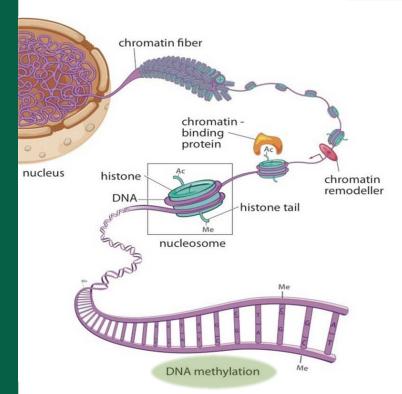


1. DNA methylation

Introduction



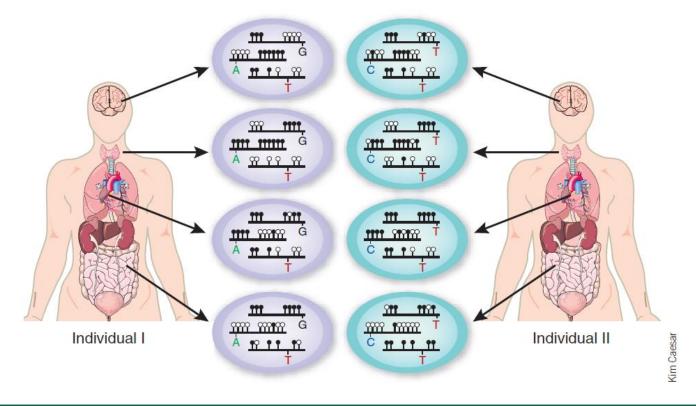
What is **DNA methylation**?



- DNA methylation is a major <u>epigenetic</u> mechanism involving direct chemical modification to the DNA
- Epigenetics is the study of the mechanisms that cause changes in gene expression but are not due to the change in the DNA sequences, including DNA methylation, histone modification, noncoding RNA expressions

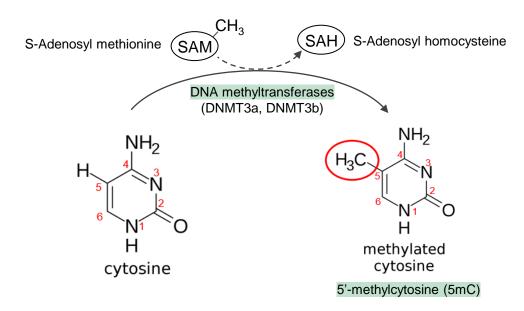


Tissue-specific **DNA methylation** and epigenetic heterogeneity among individuals





What is **DNA methylation**?



- Most common form of DNA methylation is 5-methylcytosine (5mC)
- Three types of DNA methyltransferases (DNMT's): DNMT1, DNMT3a, DNMT3b
- 5mC is inherently mutagenic: spontaneously undergo deamination, leading to C->T transitions.



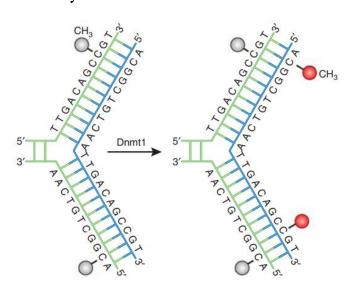
Oxidized Derivatives of 5-MethylCytosine

Cytosine Deamination

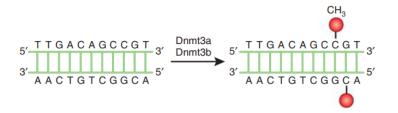


DNA methylation **enzymes**

DNMT1: maintenance DNA methyltransferase:



DNMT3a, DNMT3b: de novo DNA methyltransferase:





DNA methylation **locations**

- DNA methylation primarily happens at CpG sites.
- **CpG** dinucleotides/**sites**
 - DNA regions where a cytosine is followed by a guanine in the 5'-3' direction (5'-Cytosine-phosphate-Guanine-3')
 - CG dinucleotides present on single strand of DNA
 - <u>Under-representation</u>
- **CpG island** are regions with a high frequency of CpG sites
 - Length > 200bp
 - GC percentage > 50%
 - Observed-to-expected CpG ratio > 60%
- CpG island 'shores' are regions of comparatively low CpG density, located approximately 2 kb from CpG islands



CpG sites

GpC sites

CCCGGGTCCGGGCGGGAAGAGCCGCCTCAACGGCAGGGCCCATCCGCGA

CpG island

CCCGGGTCCGGGGGGGAAGAGCCGCCTCAACGGCAGGGCCCATCCGCGA GGCTGCTCCTCCGGGCCCTGCACCGCCCTCCTGCTACTTGGACCGCTTC CTCA CGCCCTTCTCCA CCCCGCGCGCCA GCCTCCCGCGCGCA GCGTGGGG ATCTCGGCCAATAAAGGAGAAAGGGCGCGGCCCGTACGCGCCCAGGTGC GTGGGCGAGACCAGCTCACGCCCCTCCTCCAGCCGCCAAGGCCCCGGCCC CTTGACTCGCACTTTTGTCCGGTTCGAACGTTCTGCTCAGTGGTGCGTGG AATGCGAGCGCGTCTTAAAATCGATGGCGCCTAGGAGTCCATGAAATACG GTACAGGCTTCCGGCGACGGATGCCCCGCCCCTCACCCACGCTCCGCCCT CCGGGGATGCCCCACCCCTCGTGGCGGTCCCGCCCGTCCCCGCGCAGGCG CGCTCGGGCTGCCGCTGGCTCTTCGCACGCGGCCATGGCCGACTCCGAGC TGCAGCTGGTTGAGCAGCGGATCCGCAGCTTCCCCGACTTCCCCACCCCA GGCGTGGTATTCAGGTGCACGCACAGGCCGCCCTCGTGGCGCCCCGACCT CGGGAACCCTCGTCTTTCGCCCCCGGGGCCCTGCCTCCTTCGGCCCCGG CGTCACCAGGCCTGTCCTTGGGTCCAGGGACATCTCGCCCGTCCTGAAGG ACCCCCCCTCCTTCCCCCCCCCCCATCCCCCCTCCTGGCCCCACACCCTGAAG GCGACCCACGGGGGCCGCATCGACTACATCGCAGGCGAGTGCCCAGTGGC CGCATCTA GGGCGCTTCCGCCTCTGCGCGCGCCGA GGGCAGCACGTGGGC TCTGCGCGTCTGCTTGGGGGAGGGCCTTTGGGGTGCTTCAGGGGGCGCCG GGACGGGCCCTTGCTTGGGTCGCCCGGGAAGGGTTGTGAGATTGAGCCC CCGA GGCCGCCGCTGTGCA GGCGTCCTTCCCGCA GGTTCCCGGGTCCCC

TGCCCAAGCTGAATCCACAGGGCCCAGCTGCCTTGCTTCTTGTTCCTTCT GCGA GCTGGTA TTGAGCGCCTGCCACGA GCCAGGCCTTCCCTGGTGA A GA TCACGGAATGCCCACCCAGGGAAGGGCCTGGAGGCCTCCGGGAGAGAGC CCAAGAGGGGGCCCAGGGAGAACAGAGTGTTCCTGGCCGTCTTGCCTCTC CTAGGGTGTGACAGCCCACTCCCTGGACACTGCCCTGAGGAAAGCGCCAG CTCTTGCTGGAGCCACAACACTGCCAGAGCTCCCTTCTCACCTCCTGCAG GAAGCCCTCCTGACCTCCTGCCAGGCCGGGGCAGGGTTTCCCTGAGCGT CCCCCA A CCA TCA CAGCTCAGGCCA CCTCGA GA GA CTCCCTTTTTA GA CA GAAGCCCTGGTGCAGAGCTGCCTTTGAGAGTAAGCTGAGGCCTGTCAGGT TGACTCCCCTA GGAACA CA CA GCTA AGAA GTGGTCCCTTA A A AGA CA GAC CCAGGTCTGCACTCTGACCTGGAAGCAGCTCCGGGTAGGTGATGGGTAAC ATTCCTTAAATGGTGCATGTCACTGGCCTTTCAGCTGGGAGCCAACCAGG TACCCCTTGCCACCGGCCAACCCTGGCCCCTGGGGATTCCCATGCTGCCG AGTCACTCCTGTCACTTACCCTGACAGGCCTAGACTCCCGAGGCTTCCTC TTTGGCCCCTCCCTGGCCCAGGAGCTTGGACTGGGCTGCTGCTCATCCG AAAGCGGGGAAGCTGCCAGGCCCCACTCTGTGGGCCTCCTATTCCCTGG AGTA CGGGAA GGTA AGA GGGCTGGGGTGGCCAGA GGAA GGGCAGGGCCAG GCCACCGTGGCCACTCTCCCCCAGTTCTAAAAGGCCTTCCCAGGCGTGTC Under-representation: Aggregascreatestrackersessascreakes ACATAGGCTGGGCTCACACAGCCAGGTAACAGCAAGGTGGGGTTGGAGTC AGGGTCTAGGGTGGCAGCTGCCAAGCTGTGCAACAAGCTGTTTTCTGCG GGAGGCTGAGGACCACACACCACTTCCCACTCCAGGCTGAGCTGGAGATT CAGAAAGACGCCCTGGAGCCAGGACAGAGGGTGGTCGTCGTGGATGATCT GCTGGCCACTGGTGGTAAGGGTCTCCCCGCAGCCAACTGCTGTGGCTCCA AGGGCCTGGTGGGAGTGGGACAGGACCTCGCTGTGTGACATGGGATGCAG CTTACTGTTGTCCAGAGGGTGCCTGGTGGCCAGGCCGACACCTTCCTCTC CCCATGCCTTCCCCCAACCCAGGGGCTGGCCTGGAGCACCTGCTCT CTGCAGCCCAGGCCAACTGGGGACCTCACCCTCCCATCCCCAGGAACCAT GAACGCTGCCTGTGAGCTGCTGGGCCGCCTGCAGGCTGAGGTCCTGGAGT GCCTGAGCCTGGTGGAGCTGACCTCCCTTAAGGGCAGGGAGAAGCTGGCA CCTGTACCCTTCTCTCTCTCCTGCAGTATGAGTCACCACAGGGCCTCCC AGCCCAACATCTCCAGCTGGATCCCAGGGAAATATCAGCCTTGGGCAACT GCAGTGA CCAGGGGCACCGGCTGCCCACAGGGAACACATTCCTTTGCTGG GGTTCAGCGCCTCTCCTGGGGCTGGAAGTGCCAAAGCCTGGGGCAAAGCT GTGTTTCAGCCACACTGAACCCAATTACACACAGCGGGAGAACGCAGTAA

ACAGCTTTCCCAC

GA GGCCA GCGCCCCGGCCGGTCCA GCCCA GGCCCGCCGCCTCCGCCCTG GGCTGCTCCCTCCGGGCCCTGCACCGCCCTCCTGCTACTTGGACCGCTTC CTCACGCCCTTCTCCACCCCGCGCGCCAGCCTCCCGCGCGCAGCGTGGGG ATCTCGGCCAATAAAGGAGAAAGGGCGCGCGCCCGTACGCGCGCCCAGGTGC GTGGGCGAGACCAGCTCACGCCCCTCCTCCAGCCGCCAAGGCCCCCGGCCC A CAGCTGCCTGGCTGCA GTCA GAAGCGTAGCCCGA GA CAAGGAAGGGCGC AATGCGAGCGCGTCTTAAAATCGATGGCGCCTAGGAGTCCATGAAATACG GTACA GGCTTCCGGCGA CGGATGCCCCGCCCCTCACCCACGCTCCGCCCT CCGGGGATGCCCACCCTCGTGGCGGTCCCGCCCGTCCCCGCGCAGGCG CGCTCGGGCTGCCGCTGGCTCTTCGCACGCGGCCATGGCCGACTCCGAGC TGCA GCTGGTTGAGCAGCGGATCCGCAGCTTCCCCGACTTCCCCACCCCA GGCGTGGTATTCAGGTGCACGCACAGGCCGCCCTCGTGGCGCCCCCGACCT CGGGAACCCTCGTCTTTCGCCCCCGGGGCCCTGCCCTCCTTCGGCCCCGG CGTCACCAGGCCTGTCCTTGGGTCCAGGGACATCTCGCCCGTCCTGAAGG A COCCGCCTCCTTCCGCGCCCGCCATCGGCCCTCCTGGCGCGCGACACCTGAAG GCGA CCCA CGGGGGCCGCATCGA CTACATCGCA GGCGA GTGCCCAGTGGC CGCATCTAGGGCGCTTCCGCCTCTGCGCGCGCCGAGGGCAGCACGTGGGC TCTGCGCGTCTGCTTGGGGGAGGCCCTTTGGGGTGCTTCAGGGGGCGCCC GGACGGGCGCCGTGCTTGGGTCGCCCGGGAAGGGTTGTGAGATTGAGCCC CCGAGGCCGCCGCTGTGCAGGCGTCCTTCCCGCAGGTTCCGGGTCCCC Adcccaggacagdcgtgaccgagttdccgggtcagttggtctccctggag TGCCCAAGCTGAATCCACAGGGCCCAGCTGCCTTGCTTCTTGTTCCTTCT GCGAGCTGGTATTGAGCGCCTGCCACGAGCCAGGCCTTCCCTGGTGAAGA TCACGGAATGCCCACCCAGGGAAGGGAGGCCTGGAGGCCTCCGGGAGAGC CCAAGAGGTGGCCCAGGGAGAACAGAGTGTTCCTGGCCGTCTTGCCTCTC CTAGGGTGTGACAGCCCACTCCCTGGACACTGCCCTGAGGAAAGCGCCAG CTCTTGCTGGAGCCACACACTGCCAGAGCTCCCTTCTCACCTCCTGCAG GAAGCCCTCCTGACCTCCTGCCAGGCCGGGGCAGGGTTTCCCTGAGCGT CCCCCAA CCA TCACA GCTCA GGCCA CCTCGA GA GA CTCCCTTTTTA GA CA GAAGCCCTGGTGCA GAGCTGCCTTTGAGAGTAAGCTGAGGCCTGTCAGGT CCAGGTCTGCACTCTGACCTGGAAGCAGCTCCGGGTAGGTGATGGGTAAC ATTCCTTAAATGGTGCATGTCACTGGCCTTTCAGCTGGGAGCCAACCAGG TACCCCTTGCCACCGGCCAACCCTGGCCCCTGGGGATTCCCATGCTGCCG AGTCACTCCTGTCACTTACCCTGACAGGCCTAGACTCCCGAGGCTTCCTC TTTGGCCCCTCCCTGGCCCAGGAGCTTGGACTGGGCTGCTCATCCG AAAGCGGGGGAAGCTGCCAGGCCCCACTCTGTGGGCCTCCTATTCCCTGG AGTA CGGGAA GGTAA GA GGGCTGGGGTGGCCAGA GGAA GGGCA GGGCCAG GCCACCGTGGCCACTCTCCCCCAGTTCTAAAAGGCCTTCCCAGGCGTGTC AAGTGGAGCTGCTGTGGTTACAGTGGCCTTGGGAGCTCAGAGAGGTTGAG ACATAGGCTGGGCTCACACAGCCAGGTAACAGCAAGGTGGGGTTGGAGTC AGGGTCTAGGGTGGCAGCTGCCAAGCTGTGCAACAAGCTGTTTTCTGCG GGAGGCTGAGGACCACACACCACTTCCCACTCCAGGCTGAGCTGGAGATT CAGAAAGACGCCCTGGAGCCAGGACAGAGGGTGGTCGTCGTGGATGATCT GCTGGCCACTGGTGGTAAGGGTCTCCCCGCAGCCAACTGCTGTGGCTCCA AGGCCTGGTGGGAGTGGGACAGGACCTCCTGTGTGACATGGGATCCAC CTTA CTGTTGTCCA GAGGGTCCTGGTGCCAGCCCGA CACCTTCCTCTC CCCATGCCTTCCCCTACCCAGGGGCTGGCCTGGAGCACCTGCTCT CTGCAGCCCAGGCCAACTGGGGACCTCACCCTCCCATCCCCAGGAACCAT GAACGCTGCCTGTGAGCTGCTGGGCCGCCTGCAGGCTGAGGTCCTGGAGT GCGTGAGCCTGGTGGAGCTGACCTCGCTTAAGGGCAGGGAGAAGCTGGCA CCTGTACCCTTCTCTCTCTCCTGCAGTATGAGTCACCACAGGGCCTCCC AGCCCAA CATCTCCAGCTGGATCCCAGGGAAATATCAGCCTTGGGCAACT GCAGTGACCAGGGGCACCGGCTGCCCACAGGGAACACATTCCTTTGCTGG GGTTCAGCGCCTCTCCTGGGGCTGGAAGTGCCAAAGCCTGGGGCAAAGCT GTGTTTCAGCCACACTGAACCCAATTACACACAGCGGGAGAACGCAGTAA ACAGCTTTCCCAC

In mammalian genome, 70-80% CpGs are methylated.

60% of mammalian promoters have CpG islands, but they are mostly unmethylated

CG suppression



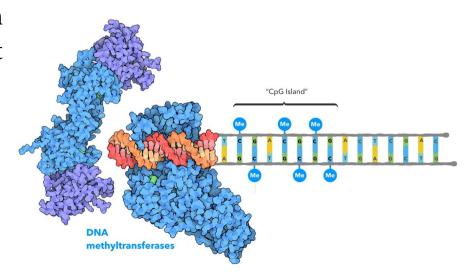
DNA methylation **functions** vs. **locations**

- DNA methylation effects on transcriptional regulation differ depending on the location of the CpG site (intragenic vs promoter region vs enhancer).
 - **Methylated** CpG island **promoters** are associated with **gene repression**.
 - CpG islands, occur and commonly span promoters of <u>house-keeping genes</u>.
 These promoter CpG islands typically remain unmethylated, resulting in active gene expression
 - Gene bodies tend to have intermediate CpG densities. Unlike CpG island promoters, extensive exonic or genic methylation is typically associated with active gene expression.
- Beyond these regions, the genome has a lower-than-expected frequency of CpG sites which are typically methylated



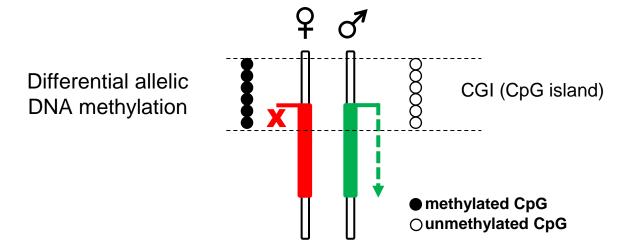
DNA methylation **functions**

- DNA methylation has been shown to be important for:
 - Genomic imprinting
 - Transposable element silencing
 - Stem cell differentiation
 - Embryonic development
 - Inflammation
 - Cancer





Imprinted Genes: mono-allelic expression



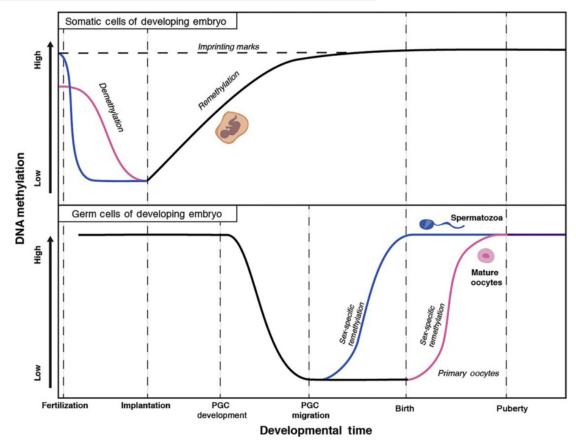
Imprinted Genes: Mono-allelic expression with parent-of-origin specificity. Have key roles in energy metabolism, placenta functions.



DNA methylation is reset during reprogramming

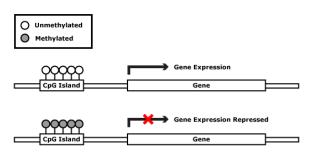
The genome undergoes two waves of global demethylation and re-methylation for the purpose of producing the next generation:

- 1. After fertilization
- 2. Germ cell development





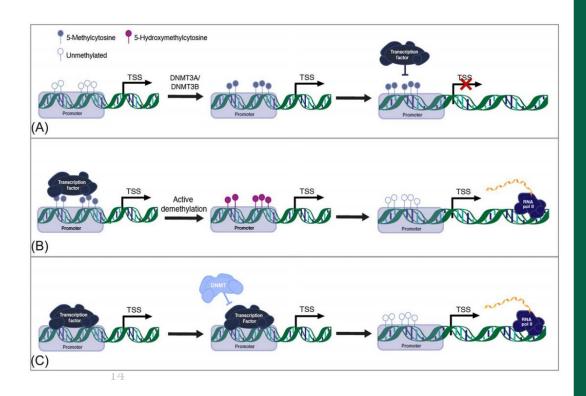
DNA methylation in gene **promoter**



Silencing of gene expression

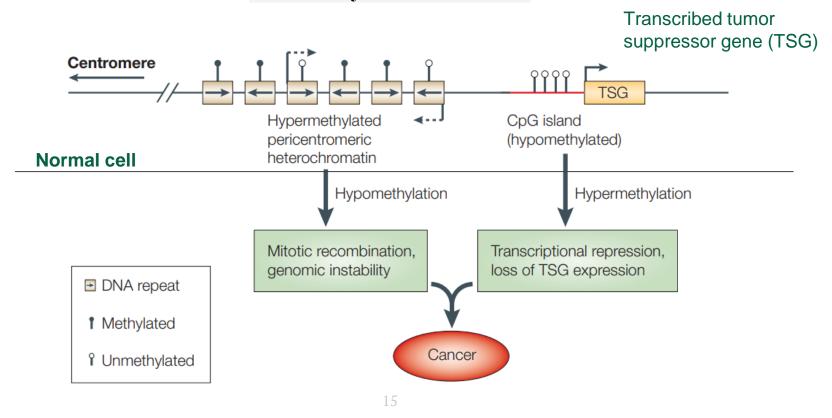
Two possible ways:

- Prevent TF from binding
- Wrap DNA up make it inaccessible.





DNA methylation and **cancer**



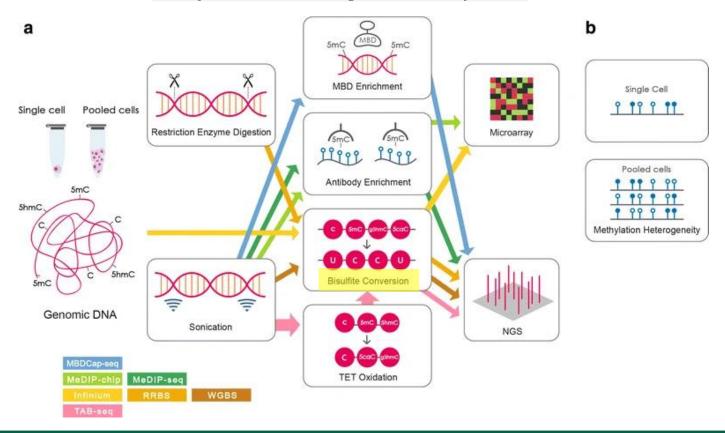


2. How to measure DNA methylation?

Bisulfite coversion



Assays for measuring DNA methylation





Bisulfite treatment

• Treatment of DNA with **bisulfite** converts cytosine residues to uracil but leaves <u>5-methylcytosine</u> residues <u>unaffected</u>.



Measuring DNA methylation



- Methylated (M)
- T Unmethylated (U)

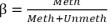
$$%Methylation = \frac{M}{M + U} \times 100$$

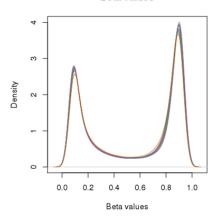


Measuring DNA methylation

For **microarrays**, there are other measurements:

$$\beta = \frac{\textit{Meth}}{\textit{Meth} + \textit{Unmeth}}$$

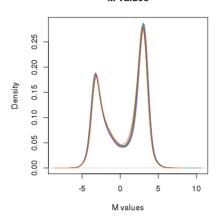




Intuitive, easy to interpret, great for visualisation

$$M = \log_2 \frac{Meth}{Unmeth}$$

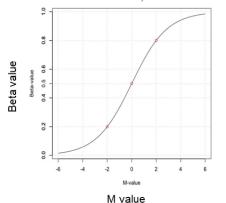
M-values



Better statistical properties, recommended for statistical testing

$$M = \log_2 \frac{\beta}{1 - \beta}$$

Du et al. 2011, BMC Bioinformatics



Can convert between them via a logit transformation



Illumina Infinium **Human** Methylation BeadChips

27k array (2009)



1 chip = 12 samples

>27,000 unique CpG sites measured in each sample

450k array (2011)



1 chip = 12 samples

>450,000 unique CpG sites measured in each sample

EPIC array (Today)



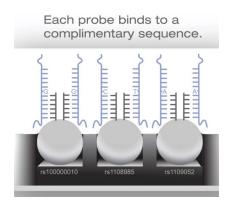
1 chip = 8 samples

>850,000 unique CpG sites measured in each sample

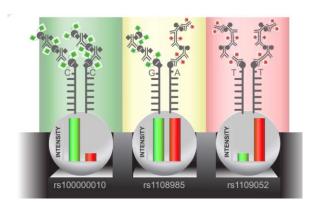
21



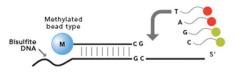
Assays for measuring DNA methylation – bisulfite microarrays

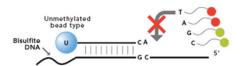


Illumina **Infinium**Methylation Assay

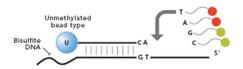


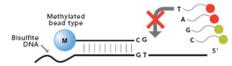






Unmethylated DNA Locus

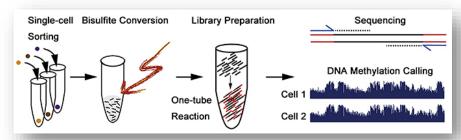






Whole-genome bisulfite sequencing (WGBS)

- WGBS experiment steps:
 - Sonication
 - End-repair
 - A-tailing
 - Adapter ligation
 - Bisulfite conversion
 - Amplification
 - Sequencing



- Advantages:
 - measure single-cytosine methylation levels genomewide
 - directly estimate the ratio of molecules methylated rather than enrichment levels



3. WGBS data analysis workflow

The real deal



WGBS data analysis steps

1. Quality control 2. Bisulfite alignment 3. Methylation Quantification 4. Visualization 5. DMRs detection

1. Quality control

Quality trimming, adapter trimming

2. Bisulfite sequence alignment

Two strategies: wild-card alignment, three letter alignment

3. Quantification of DNA methylation (methylation calling)

Sequence deduplication, .bed file format

- 4. Visualization
- 5. Differentially methylated regions (DMRs) detection



Why do we need quality control?

NGS generates highly accurate data, but it can have certain types of errors:

- contamination with adapters
- technical duplication in the library
- failure at specific parts of the flowcell
- PCR duplicates

Reads without proper trimming reads may result in

- low mapping efficiency
- mis-alignments
- errors in methylation calls since adapters are methylated
- base-call errors tend toward 50%

Tools to use: FastQC, Trim Galore!, Trimmomatics ...

3. Methylation Quantification





Sequencing data files - FASTQ

Pair-end reads sample 1

sample_1_R1_001.fastq

sample_1_R2_001.fastq

sample_2_R1_001.fastq

sample_2_R2_001.fastq

FASTQ Format

entifier ——— @HWI-EAS209 0006 FC706VJ:5:58:5894:21141#ATCACG/1

+ sign & identifier +HWI-EAS209_0006_FC706VJ:5:58:5894:21141#ATCACG/1

Base T phred Quality 1 = 29

0.7



Q-score: "Phred quality scores"

These scores represent the likelihood of the base being called wrong.

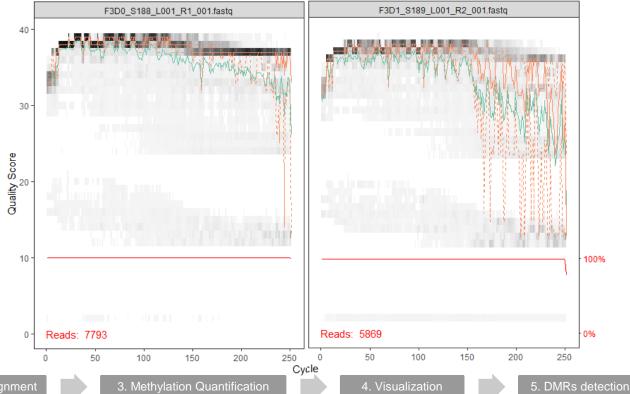
$$Q_{\rm phred} = -10\log_{10}e,$$

e is the probability that the base is called wrong.

When
$$Q = 30$$
, $e = 0.1\%$

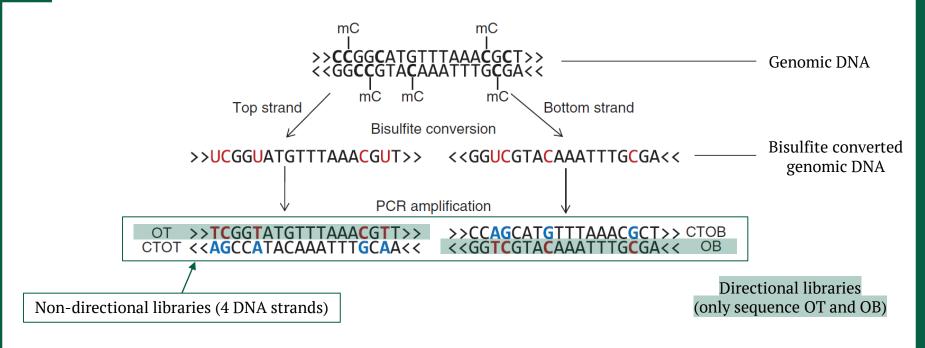
For Illumina, Q>30, the base call quality value is good enough

Quality control - FASTQC





Effect of bisulfite treatment of DNA



1. Quality control

2. Bisulfite alignment

3. Methylation Quantification

4. Visualization



Bisulfite sequence alignment – two strategies

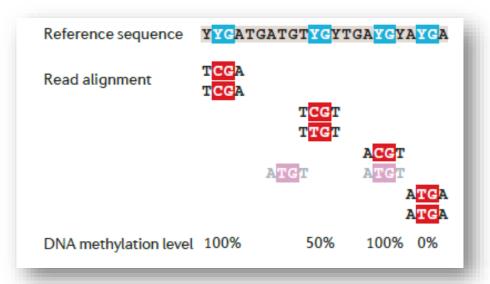
Genomic DNA sequence DNA methylation level

Bisulphite-sequencing reads ACGT, ATGA, ATGA, ATGT, TCGA, TCGA, TCGT, TTGT

1.Wild-card aligner

Replace Cs in the genomic DNA sequence (reference) by the wild-card letter Y, which matches both Cs and Ts in the read sequence

The **DNA methylation level**: the percentage of aligning Cs among all uniquely mapped reads



. Quality control

2. Bisulfite alignment



4. Visualization





Bisulfite sequence alignment – two strategies

Genomic DNA sequence DNA methylation level

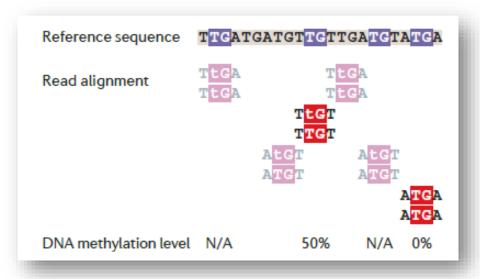
Bisulphite-sequencing reads ACGT, ATGA, ATGA, ATGT,
TCGA, TCGA, TCGT, TTGT

2. Three-letter aligner

Converting all Cs into Ts in the reads (in lower case t) and genomic DNA sequences

Carry out the alignment on a three-letter alphabet (A, G and T) using standard aligner, such as Bowtie.

The **DNA methylation level**: the percentage of aligning Cs among all uniquely mapped reads



. Quality control

2. Bisulfite alignment

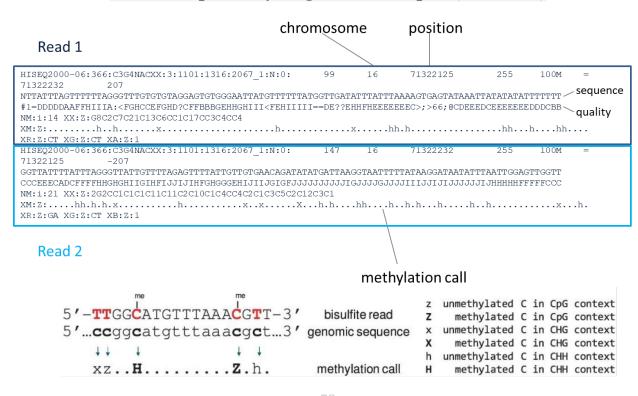


4. Visualization





Bismark primary alignment output (BAM file)



. Quality contro



3. Methylation Quantification

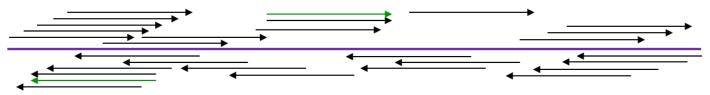


4. Visualization



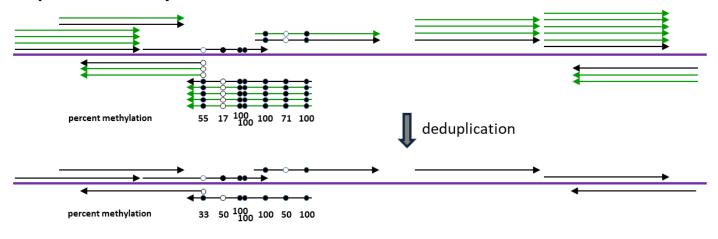


Complex/diverse library:



Sequence duplication

Duplicated library:



Quality contro

2. Bisulfite alignment

3. Methylation Quantification

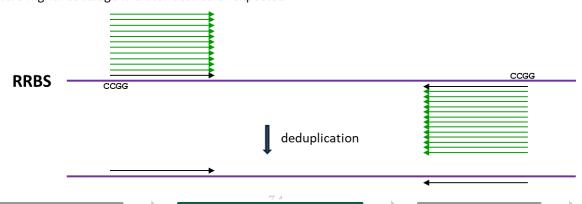
4. Visualization



Sequence Deduplication - considerations



NOT advisable for RRBS or other target enrichment methods where higher coverage is either desired or expected



1. Quality control

2. Bisulfite alignment

Methylation Quantification

4. Visualization



Methylation extraction – output .bed/bedGraph file

.bed/bedgraph file format (UCSC)

Following the track definition line are the track data in four column BED format:

```
chromA chromStartA chromEndA dataValueA
chromB chromStartB chromEndB dataValueB
```

Example of methylation extraction output (bedGraph/coverage output):

```
1 5705370 5705370 100 1 0

1 5706335 5706335 60 3 2

1 5706336 5706336 100 3 0

1 5706453 5706453 75 3 1

1 5706454 5706454 0 0 2

1 5706845 5706845 71.4285714285714 5 2

1 5706846 5706846 66.666666666667 2 1

1 5707925 5707925 0 0 1

1 5707926 5707926 66.666666666667 2 1

1 5709177 5709177 100 2 0

1 5709178 5709178 0 0 1

1 5710030 5710030 66.666666666667 4 2
```

Chromosome start end Methylation meth unmeth



Decide early on which data to use

Methylation contexts

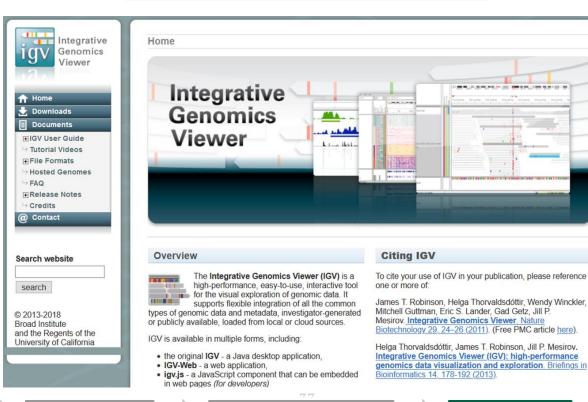
- CpG: Only generally relevant context for mammals
- CHG: Only known to be relevant in plants
- CHH: Generally unmethylated

Methylation strands

- CpG methylation is generally symmetric
- Normally makes sense to merge OT / OB strands



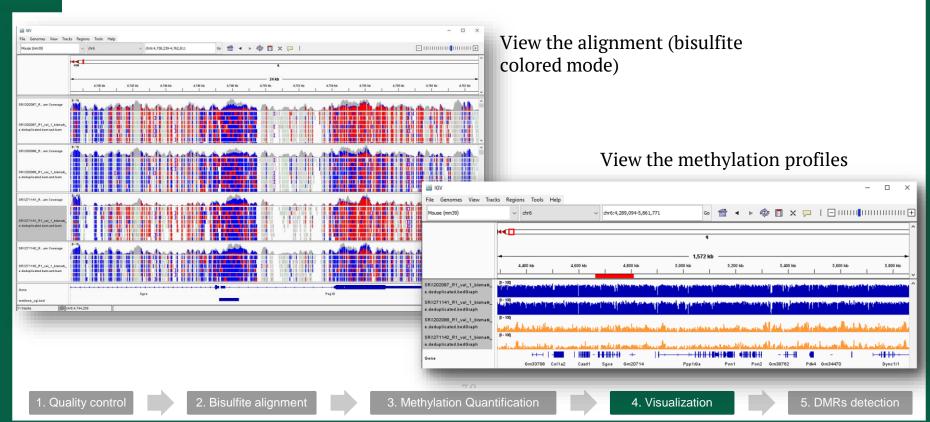
Always start by looking at your data ...



4. Visualization



Visualization of DNA Methylation: IGV





Differentially Methylated Regions (DMRs)

- **DMRs**: regions that exhibit consistently different DNA methylation levels between sample groups (e.g., cases vs. control).
- DMRs can be a single C (differentially methylated cytosine, DMC) or as large as an entire gene locus.
- Size: a few hundred to a few thousand bases



Differentially Methylated Regions (DMRs)

DMR detection:

- Basic: t-test, Wilcoxon rank-sum test
- Advanced: mixture models, Shannon entropy, feature selection, logistic M values

Test at large number of genomic loci

- Correction for multiple hypothesis testing: false discovery rate (FDR: q-value)
- Only strongest single-CpG difference tend to remain significant

To improve the statistical power

- Larger genomic regions
- Pre-selected set of candidate genomic regions

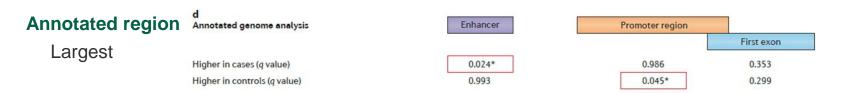
Resolution levels



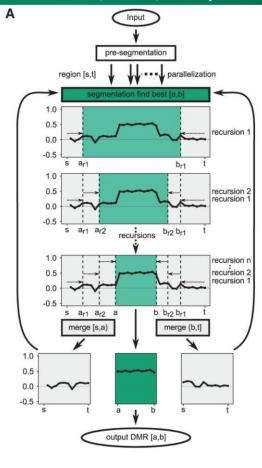
Ger	nomic DNA sequence	CG	 CG	 	co		сс	 	 CG	 CG	 CG	 CG	 CG	•••	 CG	
Cases	Sample 1 Sample 2 Sample 3	3% 2% 0%	6% 0% 1%		509 959	6	57% 74% 86%		1% 0% 2%	0% 1% 0%	1% 0% 0%	1% 0% 0%	42% 38% 41%		78% 85% 67%	
Controls	Sample 4 Sample 5 Sample 6	0% 1% 0%	2% 4% 2%		5% 139	V.	1% 2% 1%		12% 15% 19%	3% 5% 2%	15% 33% 24%	8% 11% 22%	36% 39% 33%		72% 94% 92%	

Single CpG	b Single-CpG analysis	CG1	CG2	CG3	CG4	CG5	CG6	CG7	CG8	CG9	CG10
	Higher in cases (q value)	0.333	0.993	0.085	0.068	0.993	0.993	0.993	0.993	0.196	0.993
	Higher in controls (q value)	0.993	0.732	0.993	0.993	0.070	0.104	0.104	0.110	0.993	0.351









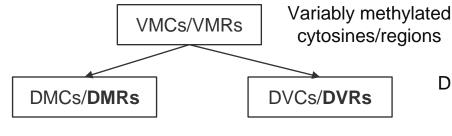
DMR detection - metilene

- The objective problem of finding DMRs has two dimensions:
 - o find a genomic region
 - the individuals of two groups are significantly distinct in their methylation levels.
- De-novo DMR detection

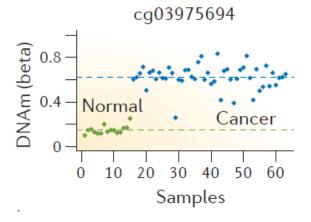


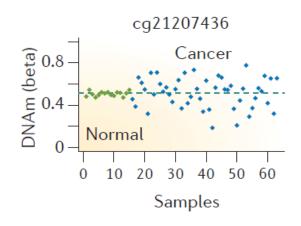
Differentially variability

Differentially methylated cytosines/regions (Most common supervised features)



Differentially variable cytosines/regions





1. Quality contro

2. Bisulfite alignment

3. Methylation Quantification

4. Visualization



The end.