



# DNA Methylation Methods and Technologies

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NBIS course - Epigenomics Data Analysis Sept 18, 2023





- i. Overview of methods for interrogation of DNA methylation
  - Overview of important concepts
  - Enrichment & targeted-based methods
  - Genome-wide methods
- ii. How to access epigenomics services for your research project at Sweden's National Genomics Infrastructure (NGI)

# **Short intro: Conversion**



Bisulfite coversion has been the "Gold standard" for DNA methylation analysis. Provides "single nucleotide resolution".



# What you need to know about bisulfite conversion

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• Very harsh chemical that degrades and fragments DNA



### New innovation- Enzymatic conversion!





Gentle with little/no strand breakage!

WGBS is the gold standard for methylome analysis, but the chemical bisulfite reaction:

- I. Damages / degrades DNA
- II. Results in fragmentation / loss
- III. Can result in CG bias and uneven genome coverage

#### **Enzymatic methylation sequencing (EM-seq)**

TET2 enzymatically oxidizes 5mC and 5hmC through a cascade reaction into 5-carboxycytosine (5caC)

5-methylcytosine (5mC)  $\rightarrow$  5-hydroxymethylcytosine (5hmC)  $\rightarrow$  5-formylcytosine (5fC)  $\rightarrow$  5-carboxycytosine (5caC)

A second enzymatic step uses APOBEC to deaminate cytosine to uracil, but does not affect 5caC.

# Base-pair resolution and quantitative measurement of methylation levels



C = methylated

U->T = unmethylated



Unmethylated CpG bases Methylated CpG bases

### Short intro: "NGS" libraries





sequencing adapters

### Short intro: "NGS" libraries





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Different approaches to reduce the genome to regions of interest (typically those with many CpG sites)

- Cost saving (less sequencing required)
- Less computationally intensive (less data generated)
- High throughput (some approaches)

## **Enrichment-based methods**



Capture of methylated DNA fragments using methylbinding protein or a anti-methyl-cytosine antibody

- MeDIP-seq (Methylated DNA immunoprecipitation):
  - ✓ Genome-wide coverage
  - $\checkmark$  ~150bp resolution.
  - ✓ Anti-body against 5-Hydroxy-methyl-cytosine
  - ✓ Relatively cost-efficient
- MBD-seq (Methylated DNA binding domain):
  - ✓ Genome-wide coverage
  - $\checkmark$  ~150bp resolution.
  - ✓ Only capture CpG methylation not CHH
  - ✓ Relatively cost-efficient



Figure from:

Lan, et al. (2011) High Resolution Detection and Analysis of CpG Dinucleotides Methylation Using MBD-Seq Technology. https://doi.org/10.1371/journal.pone.0022226

### **Enrichment-based methods**



Relative methylation level

The depth of sequence reads is taken as an indirect measurement of Methylation levels

#### Pros:

Works for different species

#### <u>Cons:</u>

- Not base-pair resolution
- Indirect measurement of DNA methylation can be more difficult to interpret
- Lab-intensive and not easily automated



# **Target-Capture**

**Target-capture** of pre-defined genomic regions, NGS library preparation, uses bisulfite conversion.

#### Pros:

- Focused set of targets regions: can achieve high coverage on target
- "Cost-effective"
- Captures millions of CpG sites (3-5M)

#### <u>Cons:</u>

- Typically only for Human, other species possible on some platforms
- Standard conversion cannot distinguish between 5mC and 5hmC



PCR & NGS

# **DNA methylation arrays**

- Bisulfite converted DNA
- >800,000 CpG sites
- 96% CpG islands
- 99% Refseq genes
- CpG sites outside of CpG islands
- Non-CpG methylated sites identified in human stem cells
- Differentially methylated sites found in cancer and several tissue types
- FANTOM 4 promoters
- DNase hypersensitive sites
- miRNA promoters



# **DNA methylation arrays**



### Pros:

- The most popular method on the market
- Base-pair resolution
- Compatible with FFPE DNA
- Compatible with 5hmC detection
- Many **R packages** available for data analysis and publically available datasets

### • <u>Cons</u>

- Human only\* (Mouse Methylation BeadChip available with 285k CpG sites or flexible iSelect Methyl Custom BeadChip, but \$\$)
- 850k out of 29M CpG sites

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		illumina 4305493023

### Reduced Representation Bisulfite Sequencing (RRBS)



- "reduces" the genome to informative regions with high CG content
- Based on restriction digestion with an enzyme that cuts at CCGG (MspI)

### **Pros:**

- Compatible with most species
- Low cost
- Base-pair resolution (bisulfite)
- reads are heavily concentrated to CpG islands
- High throughput

### Cons

- Does not capture all promoters or CpG islands
- Results can vary depending on input DNA quality / contaminants in the sample

• Mspl (C^CGG)

Methylation insensitive restriction enzymes

- Taql (T^CGA)
- Base-pair resolution



### RRBS





Figure adapted from Meissner et al. Nature 454 (2008) https://doi.org/10.1038/nature07107

highly conserved non-coding element



# Whole Genome Methylome Sequencing

- Many acronyms ; WGBS, MethylC-seq, BS-seq
- "Unbiased" no selection or enrichment
- Genome-wide coverage of all cytosines
- Base-pair resolution
- Uses bisulfite conversion or enzymatic conversion to distinguish methylated from unmethylated cytosines



### Many different approaches ...

DNA



J Nordlund, Chapter Eleven - Advances in whole genome methylomic sequencing, Epigenetics Methods, Academic Press (2020), https://doi.org/10.1016/B978-0-12-819414-0.00011-2.

### Whole Genome Methylome Sequencing



#### Direct read out of DNA modifications by single molecule, long read technologies (PacBio, Oxford Nanopore)



In theory can detect all sorts of DNA modification-Challenge is to train models to correctly detect specific modifications

# Short vs long-read sequencing, what's the difference?

### **Short-read**

#### Illumina

### Pros:

- Low cost
- High throughput
- Detect 5mC & 5hmC \*depending on library prep applied
- Species agnostic

### Cons:

- Requires conversion of (un)modificed bases DNA with chemicals or enzymes
- 5mC cannot be distingushed from 5hmC (and other types of marks) without specific conversion approaches

### Long-read

### PacBio/ONT

#### Pros:

- Base modification can be read directly from sequencing
- Maintain phasing information
- Detect 4mC, 5mC, 5hmC, 5fC, 5caC, 6mA, etc
- Species agnostic

#### Cons:

- Cost (high coverage needed) limiting for large genomes
- Difficult to detect signals
- Low throughput



Foox J, Nordlund J, et al. The SEQC2 epigenomics quality control (EpiQC) study. Genome Biol 2021: https://doi.org/10.1186/s13059-021-02529-2

#### **EPIC** arrays

• duplicate/triplicate at 3 labs

#### WGBS

- TruSeq DNA methylation (Illumina)
- Accel-NGS methyseq (Swift)
- SPLAT (Raine et al, NAR 2017)

#### OXBS

• TrueMethyl oxBS-seq (NuGEN)

#### **Enzymatic deamination**

• EM-seq (NEB)

#### **ONT:** direct methylation calling

7 cell lines

#### Alignment and methylation calling:

- BISMARK
- BitMapperBS
- BSSeeker2
- Bwa-meth
- Gem-bs

#### **Microarray normalization**

• 26 between-array and withinarray normalization methods





Overall, no major quantitative difference between pipelines but bwa-meth was easiest to implement and retained most data.

Noticeable inter- and intra-library differences







Overall, no major quantitative difference between methylation (beta-values) called after libraries were normalized for nr reads mapped (see next slide).

But they did differ in number of CpG sites detected!

Foox J, Nordlund J, et al. The SEQC2 epigenomics quality control (EpiQC) study. Genome Biol 2021: https://doi.org/10.1186/s13059-021-02529-2



Correlation in DNA methylation estimation decreases as coverage decreases r = 0.990 r = 0.980 r = 0.951 r = 0.895 r = 0.995 Total EMseq (53X) ſ 40x 30x 20x 10x 5x

Foox J, Nordlund J, et al. The SEQC2 epigenomics quality control (EpiQC) study. Genome Biol 2021: https://doi.org/10.1186/s13059-021-02529-2

# Reproduci<mark>bility & qualit</mark>y







(a) Concordance between microarray replicates across the epigenome, by normalization pipeline

None PBC SWAN RCP 6 None Median: 73% Median: 71% (Median: 87%) (Median: 89%) 4 2 6 pQuan Incompatible (Median: 85%) Median: 69% Median: 86% 4 2 methods Median 6 dasen Median: 86% Median: 70% Median: 88% Median: 89% 4 0.9 20 density 0 0 0 0 0 0 funnorm 0.8 (Median: 90%) Median: 79% (Median: 84%) Median: 90% 0.7 6 ENmix (Median: 88%) (Median: 88%) (Median: 73%) (Median: 85%) 4 2 0 0.6 6 SeSAMe Incompatible (Median: 90%) Median: 89% (Median: 89%) 4 2 methods 0 6 GMQN (Median: 73%) Median: 60% (Median: 83%) Median: 90% 4 2 0 25% 50% 75% 100%0% 25% 50% 75% 25% 50% 75% 100%0% 25% 50% 75% 100% 0% 100%0% DNA methylation variance explained by cell line

funnorm + RCP worked best on these samples

Foox J, Nordlund J, et al. The SEQC2 epigenomics quality control (EpiQC) study. Genome Biol 2021: https://doi.org/10.1186/s13059-021-02529-2

# Single-cell WGBS





### Single cell WGBS

- ✓ Single stranded library prep
- ✓ FACS sorting required (384 plates)
- ✓ Plate- based low throughput (although autmation enable troughput of >1000 cells/exp)
- ✓ Expensive
- ✓ Sparse information-At most 50% CpG sites coverage, usually a lot less



Numerious protocols exist for scWGSB, RRBS, etc – and even integrate transcriptomics in and DNA methylation measurements from the same cell!



Lee, J. et al. Single-cell multiomics: technologies and data analysis methods. *Exp Mol Med* **52**, 1428–1442 (2020). <u>https://doi.org/10.1038/s12276-020-0420-2</u>

# In summary, there are many approaches for studying DNA methylation





Gouil and Keniry, Essays in Biochemistry, 2019. <u>https://doi.org/10.1042/EBC20190027</u>

# So which method should I choose?

#### ANALYSIS

#### nature biotechnology

Comparison of sequencing-based methods to profile DNA methylation and identification of monoallelic epigenetic modifications

R Alan Harris<sup>1,\*</sup>, Ting Wang<sup>2</sup>, Cristian Coarfa<sup>1</sup>, Raman P Nagarajan<sup>3</sup>, Chibo Hong<sup>3</sup>, Sara L Downey<sup>3</sup>, Brett E Johnson<sup>3</sup>, Shaun D Fouse<sup>3</sup>, Allen Delaney<sup>4</sup>, Yongjun Zhao<sup>4</sup>, Adam Olshen<sup>3</sup>, Tracy Ballinger<sup>5</sup>, Xin Zhou<sup>2</sup>, Kevin J Forsberg<sup>2</sup>, Junchen Gu<sup>2</sup>, Lorigail Echipare<sup>6</sup>, Henriette O'Geen<sup>6</sup>, Ryan Lister<sup>7</sup>, Mattia Pelizzola<sup>7</sup>, Yuanxin Xi<sup>8</sup>, Charles B Epstein<sup>9</sup>, Bradley E Bernstein<sup>9–11</sup>, R David Hawkins<sup>12</sup>, Bing Ren<sup>12,13</sup>, Wen-Yu Chung<sup>14,15</sup>, Hongcang Gu<sup>9</sup>, Christoph Bock<sup>9,16–18</sup>, Andreas Gnirke<sup>9</sup>, Michael Q Zhang<sup>14,15</sup>, David Haussler<sup>5</sup>, Joseph R Ecker<sup>7</sup>, Wei Li<sup>8</sup>, Peggy J Farnham<sup>6</sup>, Robert A Waterland<sup>1,19</sup>, Alexander Meissner<sup>9,16,17</sup>, Marco A Marra<sup>4</sup>, Martin Hirst<sup>4</sup>, Aleksandar Milosavljevic<sup>1</sup> & Joseph F Costello<sup>3</sup>

Foox et al. Genome Biology (2021) 22:332 https://doi.org/10.1186/s13059-021-02529-2

Genome Biology

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#### RESEARCH

The SEQC2 epigenomics quality control (EpiQC) study

Jonathan Foox<sup>1,21</sup>, Jessica Nordlund<sup>3,4†</sup>, Claudia Lalancette<sup>51</sup>, Ting Gong<sup>61</sup>, Michelle Lacey<sup>7†</sup>, Samantha Lent<sup>8†</sup>, Bradley W. Langhorst<sup>6</sup>, V. K. Chaithanya Ponnalun<sup>9</sup>, Louise Williams<sup>9</sup>, Karthik Ramaswamy Padmanabhan<sup>5</sup>, Raymond Cavalcante<sup>5</sup>, Anders Lundmark<sup>3,4</sup>, Daniel Butler<sup>1</sup>, Christopher Mozsary<sup>1</sup>, Justin Gurvitch<sup>1</sup>, John M. Greally<sup>10</sup>, Masako Suzuki<sup>10</sup>, Mark Menor<sup>6</sup>, Masaki Nasu<sup>6</sup>, Alicia Alonso<sup>11</sup>, Caroline Sheridan<sup>1,11</sup>, Andreas Schere<sup>4,12</sup>, Stephen Bruinsma<sup>13</sup>, Gosia Golda<sup>14</sup>, Agata Muszynska<sup>15</sup>, Pawel Pt. Łabal<sup>15</sup>, Matthew A. Campbell<sup>9</sup>, Frank Wos<sup>16</sup>, Amanda Raine<sup>34</sup>, Urilak Lijledahl<sup>34</sup>, Tomas Axelsson<sup>34</sup>, Charles Wang<sup>17</sup>, Zhong Chen<sup>17</sup>, Zhaowei Yang<sup>17,18</sup>, Jing Ji<sup>17,18</sup>, Xiaopeng Yang<sup>16</sup>, Hongwei Wang<sup>30</sup>, Ari Melnick<sup>1</sup>, Shang Guo<sup>21</sup>, Alexander Blume<sup>22</sup>, Vedran Franke<sup>22</sup>, Inmaculada Ibanez de Caceres<sup>4,3</sup>, Carlos Rodriguez-Antolin<sup>42,3</sup>, Rocio Rosas<sup>423</sup>, Justin Wade Davis<sup>6</sup>, Jennifer Ishil<sup>16</sup>, Dalila B. Megherbi<sup>24</sup>, Wenming Xiao<sup>5</sup>, Will Liao<sup>16</sup>, Joshua Xu<sup>26</sup>, Huixiao Hong<sup>26</sup>, Baitang Ning<sup>26</sup>, Weida Tong<sup>26</sup>, Altuna Akalin<sup>22</sup>, Yunliang Wang<sup>21\*</sup>, Youping Deng<sup>6\*</sup> and Christopher E. Mason<sup>1,2,27,28</sup>

Essays in Biochemistry (2019) 63 639–648 https://doi.org/10.1042/EBC20190027



**Review Article** 

#### Latest techniques to study DNA methylation

#### 0 Quentin Gouil $^{1,2}$ and 0 Andrew Keniry $^{1,2}$

<sup>1</sup>Epigenetics and Development Division, Walter and Eliza Hall Institute of Medical Research, Parkville, Australia; <sup>2</sup>Department of Medical Biology, University of Melbourne, Parkville, Australia

### Species

- Sample availability
- DNA quality
- Scientific question(s)
- Budget

Lee et al. Experimental & Molecular Medicine (2020) 52:1428-1442 https://doi.org/10.1038/s12276-020-0420-2

#### Experimental & Molecular Medicine

#### **REVIEW ARTICLE**

#### **Open Access**

Single-cell multiomics: technologies and data analysis methods

Jeongwoo Lee<sup>1</sup>, Do Young Hyeon<sup>1</sup> and Daehee Hwang<sup>1</sup>

#### genetics

PERSPECTIVE

Single-cell and single-molecule epigenomics to uncover genome regulation at unprecedented resolution

Efrat Shema <sup>(1,2,4</sup>, Bradley E. Bernstein<sup>1,2</sup> and Jason D. Buenrostro <sup>(1,2,3\*</sup>

Yong et al. Epigenetics & Chromatin (2016) 9:26 DOI 10.1186/s13072-016-0075-3 **Epigenetics & Chromatin** 

**Open Access** 

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#### REVIEW

#### Profiling genome-wide DNA methylation

Wai-Shin Yong<sup>1†</sup>, Fei-Man Hsu<sup>2†</sup> and Pao-Yang Chen<sup>1†</sup>

### **Epigenomics services offered by the National Genomics Infrastructure (NGI)**





### NGI is a facility within the SciLifeLab Genomics Platform located at two nodes:

NGI-Uppsala

- SNP&SEQ Technology Platform (UU)
- Uppsala Genome Centre (UU)

#### NGI-Stockholm

• SciLifeLab Solna (KTH, KI, SU)

### NGI's project portal



- All projects submitted through a **common order system**
- Projects are dynamically allocated between Stockholm/Uppsala depending on type of application, queue situation, or request by researcher



# Genotyping and sequencing on all scales



#### A decade of sequencing at NGI Cumulative Bases Samples per year Cumulative gigabases Samples per year (sequencing) Year

### Statistics for 2022:

- 1000 projects / 90,000 samples
- **912 Terabases** (10<sup>12</sup>) of sequence data

As of Jan 1, 2022 NGI has delivered a total of 6.3 Petabases (10<sup>15</sup>) of sequencing data

# Support



### **Pre support**

- **Project design** via discussions with expert project coordinators
- Advise in sample collection and/or preparation
- **DNA extraction services available** for specific applications
- Sample quality (QC) for all incoming samples and user-made libraries

### **Post support**

- Control over produced data: making sure data meet our high standards in terms of quality and yield.
- Open source Bioinformatic pipelines for a wide range of applications: *NF-core lecture*
- Data delivered via UPPMAX

## **Epigenetic methods available at NGI**







Additional information about sequencing applications that NGI supports: https://ngisweden.scilifelab.se

Don't hesitate to reach out to NGI's project coordinators: support@ngisweden.se

-or mejessica.nordlund@medsci.uu.se / <u>seq@medsci.uu.se</u>