

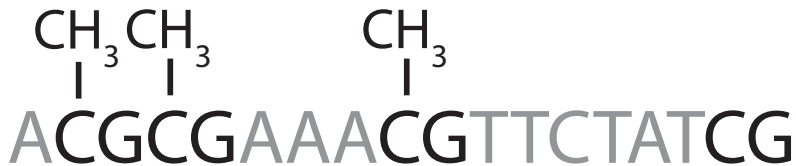
DNA methylation

ACGCGAAACGTTCTATCG

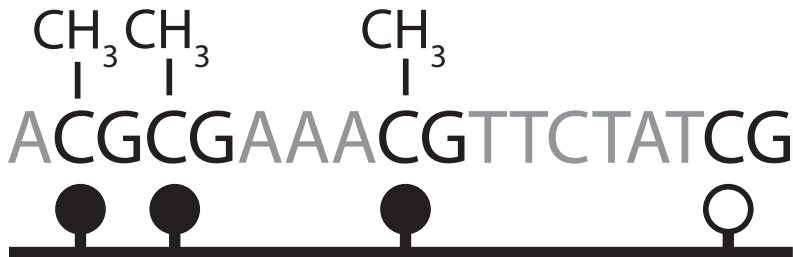
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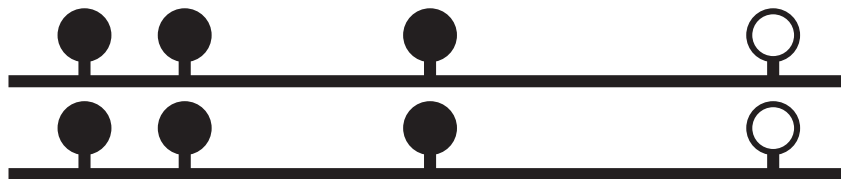
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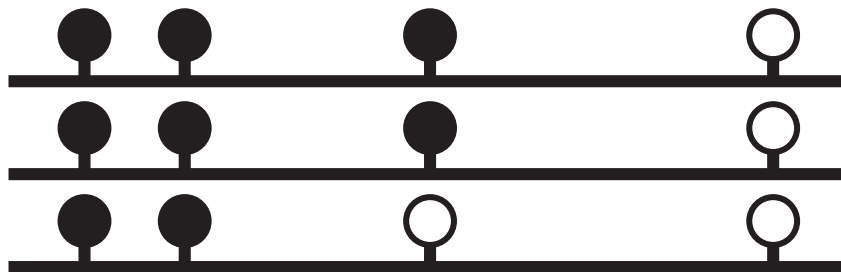
Measuring DNA methylation



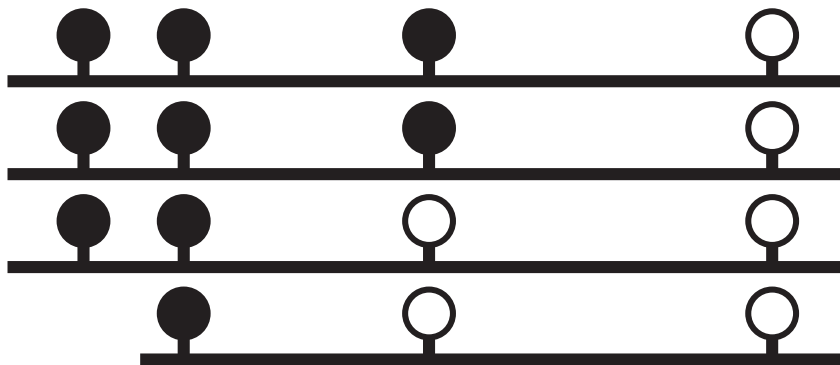
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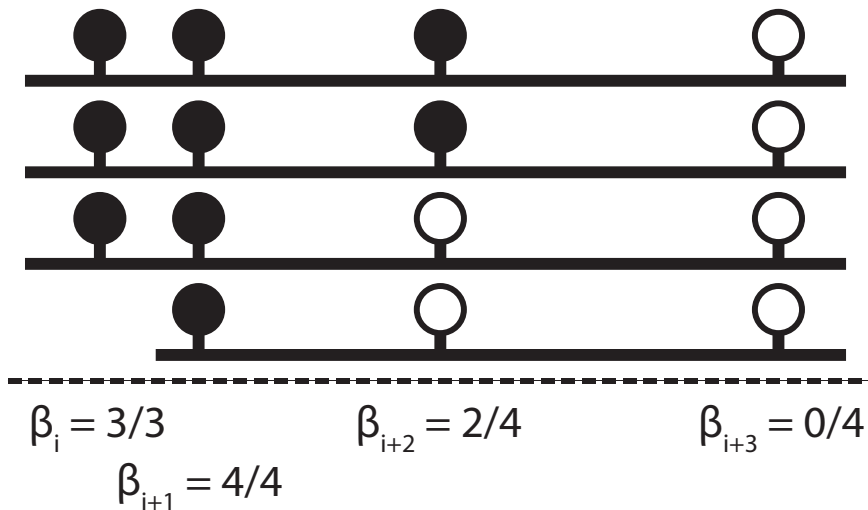
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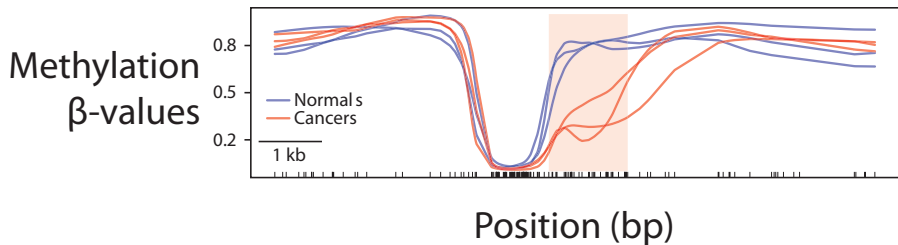
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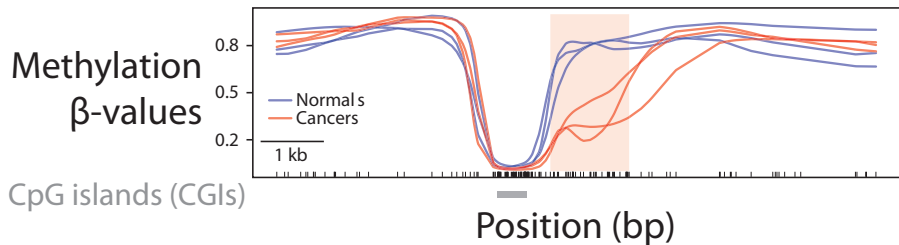


Differentially methylated regions (DMRs)¹



¹Hansen, K. D. et al. Nat Genet 43, 768–775 (2011)

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Why I care about simulating DNA methylation data

Methods development and validation

- Do methods designed to find DMRs actually work?
- What method reigns supreme?

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How to decide?

- No “gold standard” data \Rightarrow simulate
- No simulation software \Rightarrow I'm writing `methsim`.

Simulation approaches

Simulate β -values

- Simulate independent $\beta_i \stackrel{d}{=} \text{Beta}(\mu_i, \nu_i)$ + induce correlation via variogram model.

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Simulate individual methylation events

- Higher resolution.
- Contains the mechanistic dependence structure.
- **Difficult given current data.**

My solution

methsim: An R package for simulating whole genome DNA methylation data.

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Outline of `methsim`

- 1 Segment genome into “region of similarity” (MethylSeekR¹)
- 2 Simulate “meta-haplotypes” within each region using Markov model.
- 3 Simulate sequencing of reads.

^aBurger, L., Gaidatzis, D., Schübeler, D. & Stadler, M. B. *Nucleic Acids Res* (2013). doi:10.1093/nar/gkt599

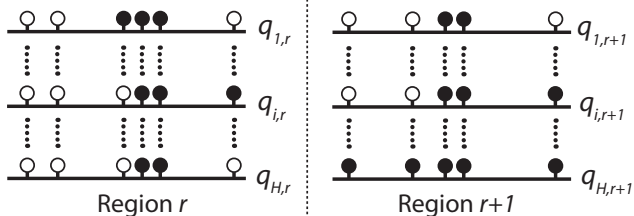
Simulating *meta-haplotypes*

(2) For each region:

Simulate each meta-haplotype using a Markov model

Transition matrices depend on distance between CGs and the type of region

Assign haplotype i in region r frequency $q_{i,r}$



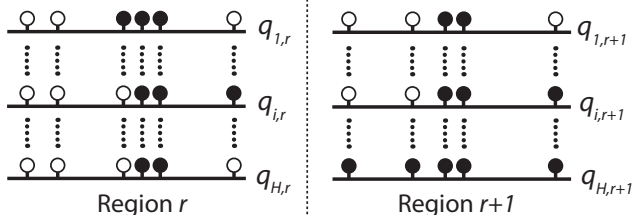
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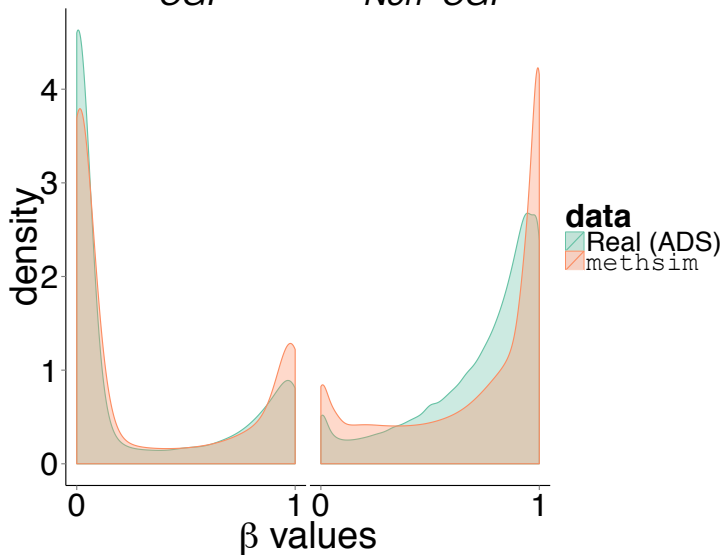


(3) Simulate read positions

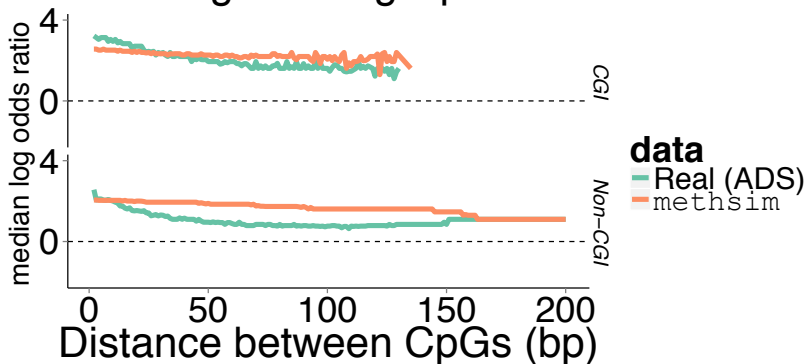
Simulate reads for region r by sampling from i^{th} haplotype with probability $q_{i,r}$

Simulate sequencing error

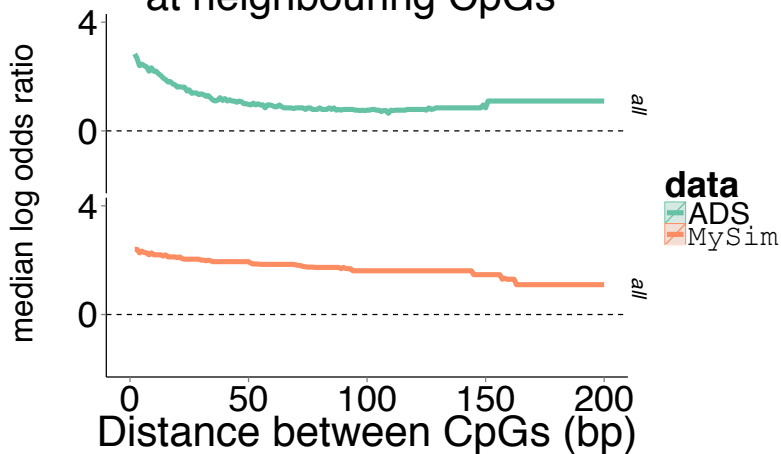
Distribution of β values



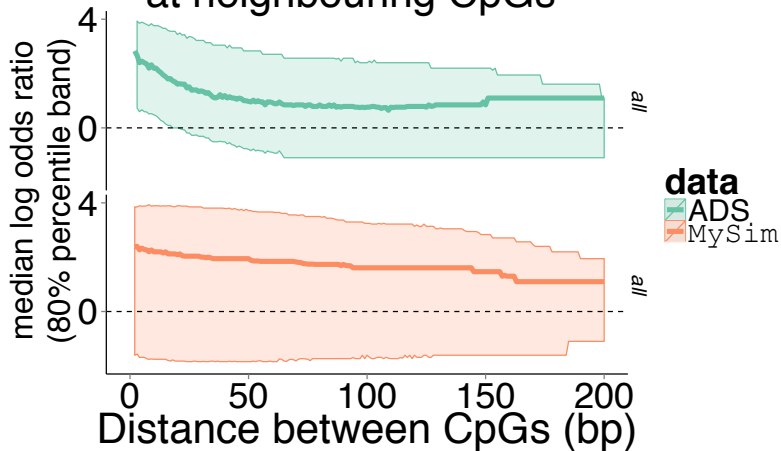
Within haplotype co-methylation at neighbouring CpGs



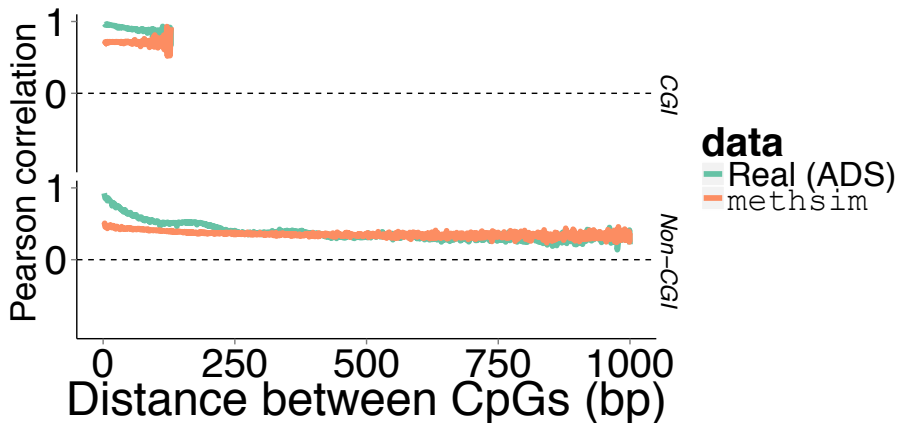
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Correlations of pairs of β values



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- Will be using `methsim` to simulate data with inserted DMRs and compare DMR-detection methods.
- `methsim` is open source and developed on GitHub.

Thanks

For advice and supervision

- Terry Speed (WEHI) and Peter Hall (University of Melbourne).

For data

- Ryan Lister (UWA).

For R and C++ help

- Bioconductor and Rcpp mailing lists, especially Martin Morgan.

For funding

- Australian Postgraduate Award, Victorian Life Sciences Computing Initiative.

For sanity

- Friends and family.

To find out more

www.peterhickey.org/ASC2014
GitHub/Twitter: @PeteHaitch