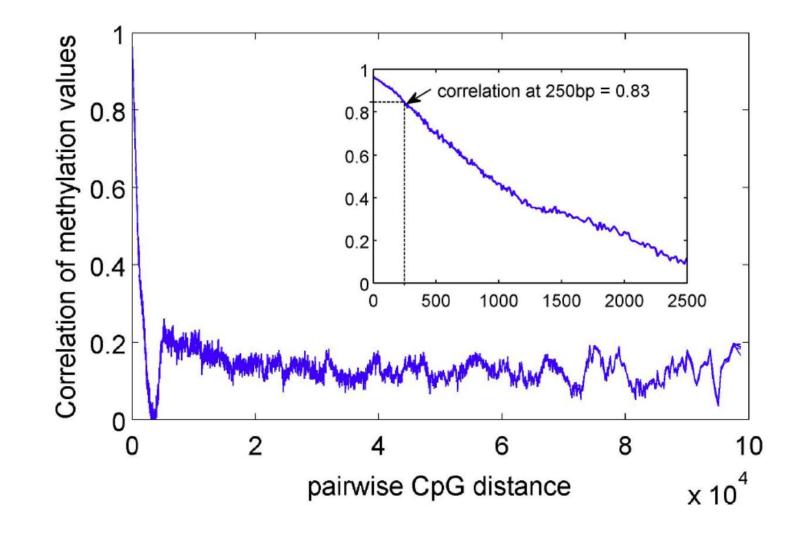
Outline for Session 4 (1.30 – 3.00pm)

- Multiple-testing
- Bumphunting
- Annotation
- Example of DNA methylation studies that are not EWAS
 - Genetic control of DNA methylation
 - Epigenetic aging

Probe correlation



Ong et al Aging Cell 2013

Multiple testing correction

- We are performing ~400,000 test per EWAS
- Expect a large number of nominally significant hits
 - 20,000 at p < 0.05
 - 400 at p < 0.001
- Need to determine an appropriate significance threshold

Multiple testing correction – "The Ugly"

- Use 5 x 10⁻⁸ because that is what we use in GWAS...
- We know the correlation structure in DNA methylation does not extend as far as SNP LD
- Probably not bad for the current generation of arrays...

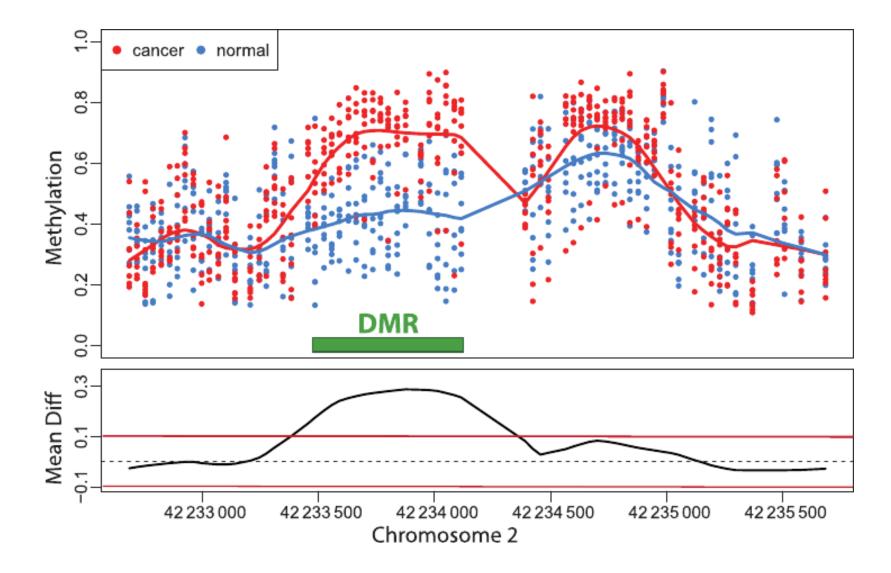
Multiple testing correction – "The Bad"

- False Discovery Rate
- DNA methylation data is correlated
- Standard FDR approaches assume the data is independent
- Results in an inflation in the FDR

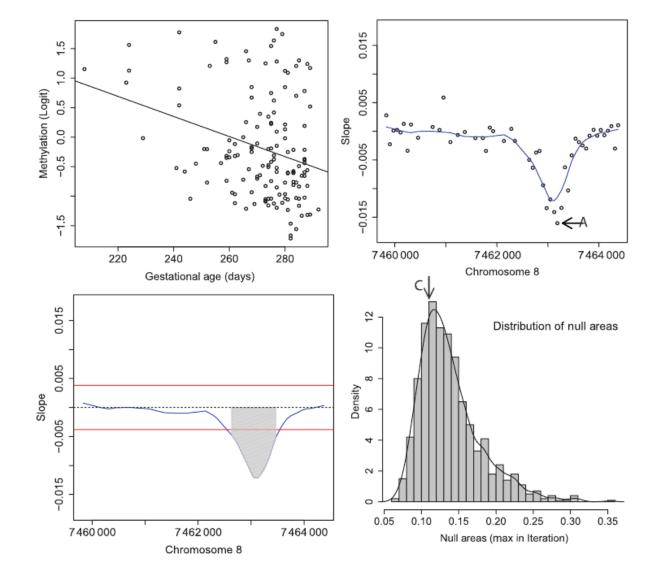
Multiple testing correction – "The Good"

- Bonferroni correction
- Divide 0.05 by the number of test performed
- Assumes all tests are independent
- Results in conservative threshold (may miss true positives)
- 450K array 450,000 tests p < 1.1 x 10⁻⁷
- EPIC array 850,000 tests p < 5 x 10⁻⁸

- Several methods have been proposed to look for "bumps" in EWAS results
- Look at combined evidence of association across multiple methylation sites
- May improve power if a region has multiple independent signals



• Bumphunter



- DMRcate
- Applies a similar approach to bumphunter
- Gives marked better results

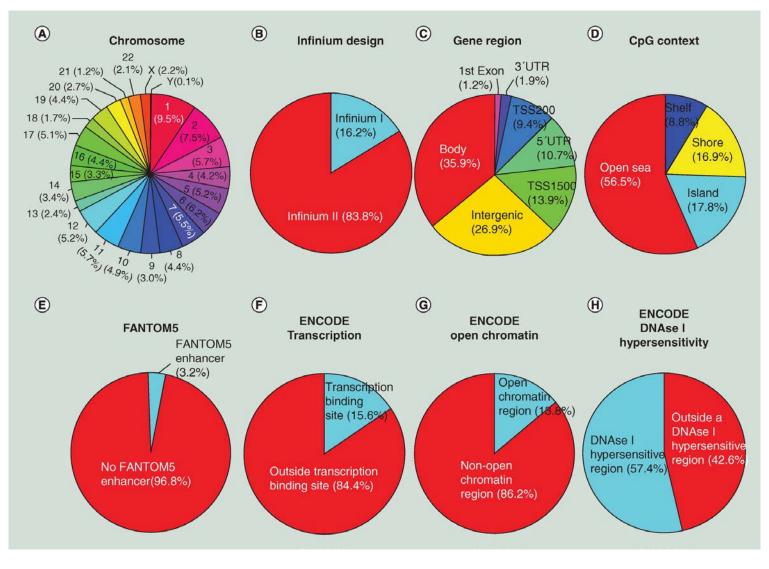
Functional Annotation

• A range of annotations have been generated for Illumina array data

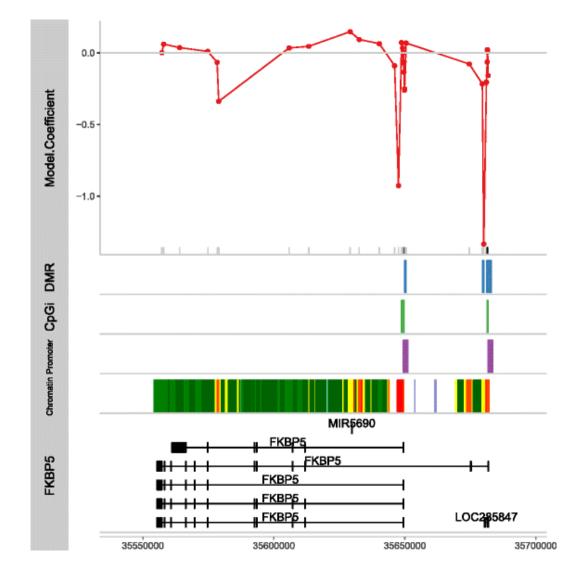
• Includes

- Chromosome, position
- Nearest genes and distance to them
- Position relative to CpG islands
- Probe sequence
- SNPs in probe binding region
- ...

Functional Annotation



Functional Annotation



Functional annotation

GREAT improves functional interpretation of *cis*-regulatory regions

Cory Y McLean¹, Dave Bristor^{1,2}, Michael Hiller², Shoa L Clarke³, Bruce T Schaar², Craig B Lowe⁴, Aaron M Wenger¹ & Gill Bejerano^{1,2}

http://bejerano.stanford.edu/great/public/html/index.php

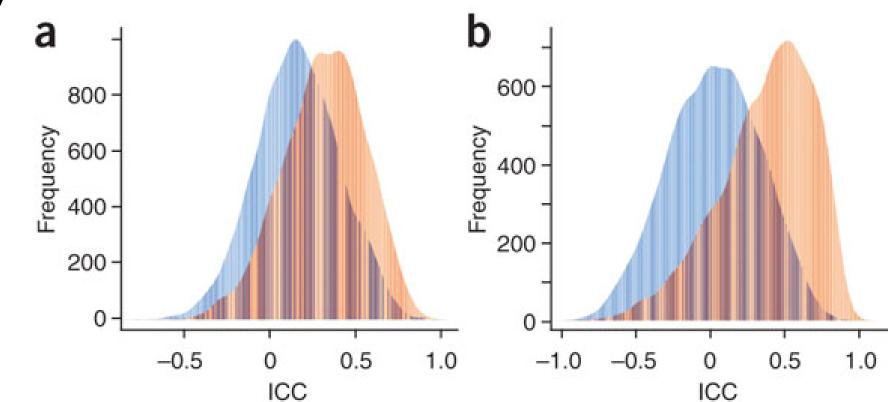
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Genetic Control of DNA methylation

- There is much interest in the transmission of DNA methylation across generations
- Potential to pass on environmental insults across generations?
- Epigenetic inheritance?
- Genetic influences on DNA methylation?

Genetic Control of DNA methylation

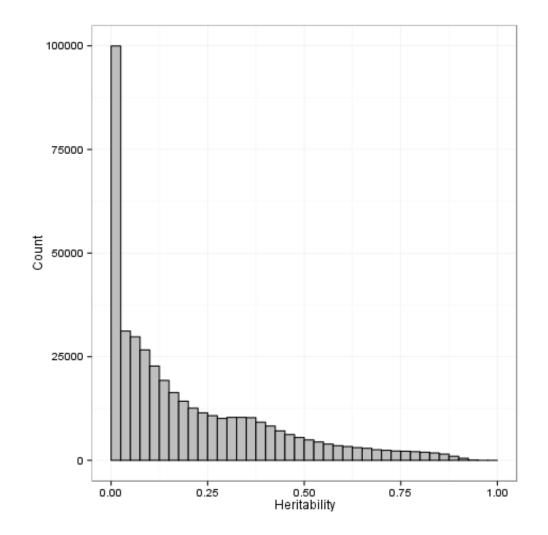
- Kamisky et al 2007
- MZ twins have have more similar DNA methylation than DZ twins
- Chorionicity has an effect

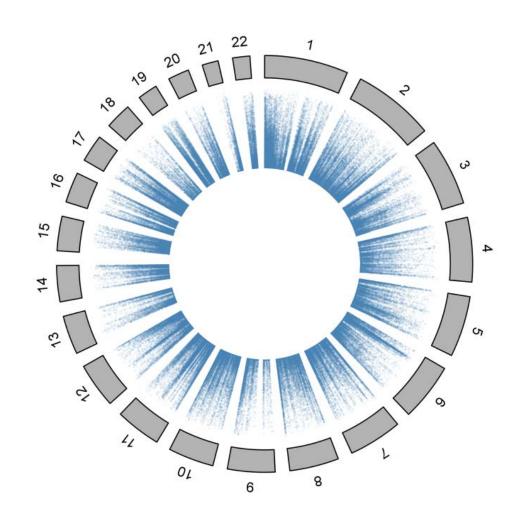


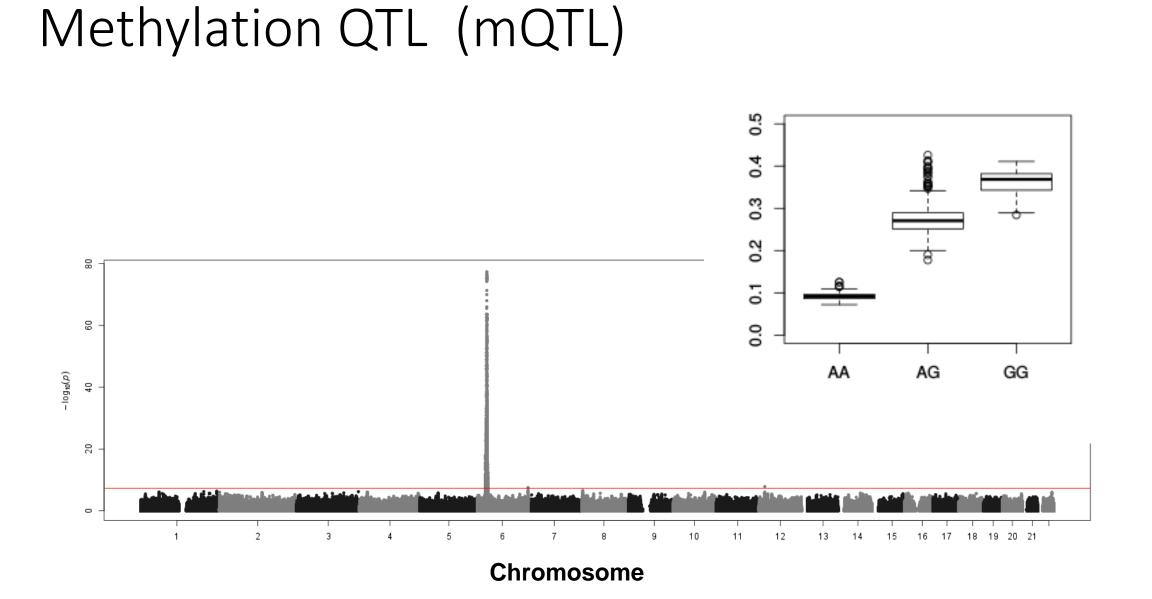
Familial Correlations of DNA methylation

Relationship	Pairs (n)	Correlation	Expected ^a		
MZ twins	67	0.200	h^2		
DZ twins	111	0.109	h²/2		
Siblings	262 ^b	0.090	h²/2		
Parent-Offspring	362 ^b	0.089	h²/2		
Mother-Offspring	190	0.097	h²/2		
Father-Offspring	172	0.085	h²/2		
Parent-Parent	58	0.023	0		
Unrelated	187,331 ^b	-0.002	0		

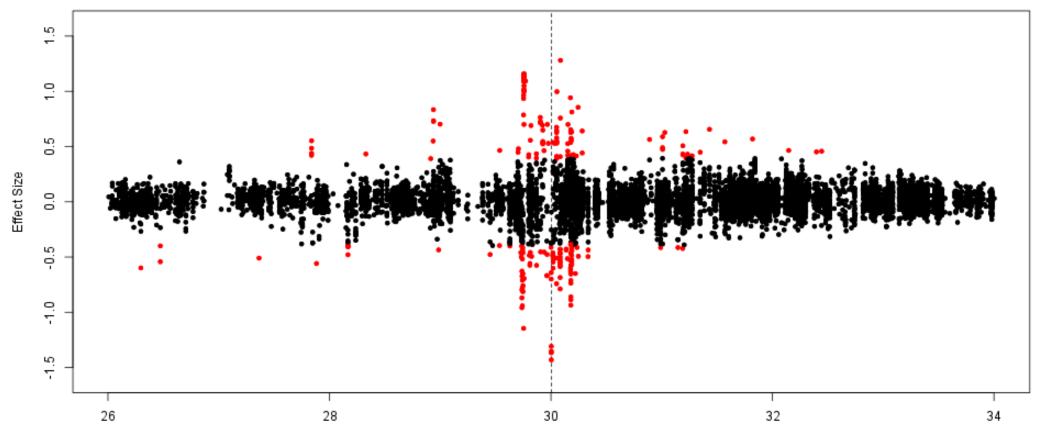
Heritability





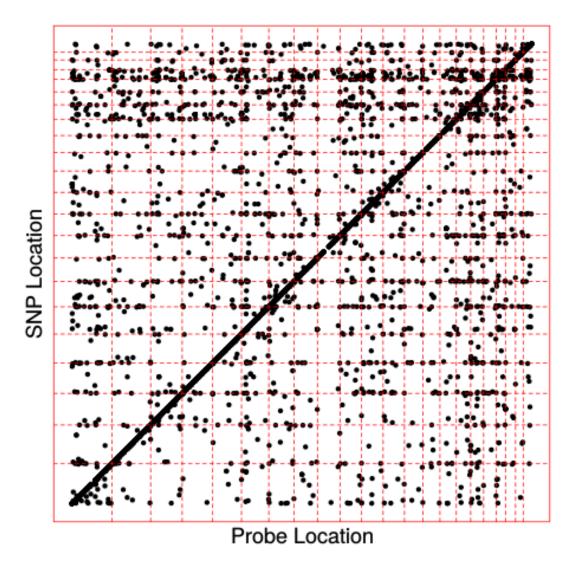


Methylation QTL



Position (Mbp)

Methylation QTL Everywhere!



Epigenetic Clock

- DNA methylation is correlated with age
 - Global change to high DNA methylation
 - Individual loci have varying amounts of change with age
- Several methods have been presented to use DNA methylation data to make a predictor of age

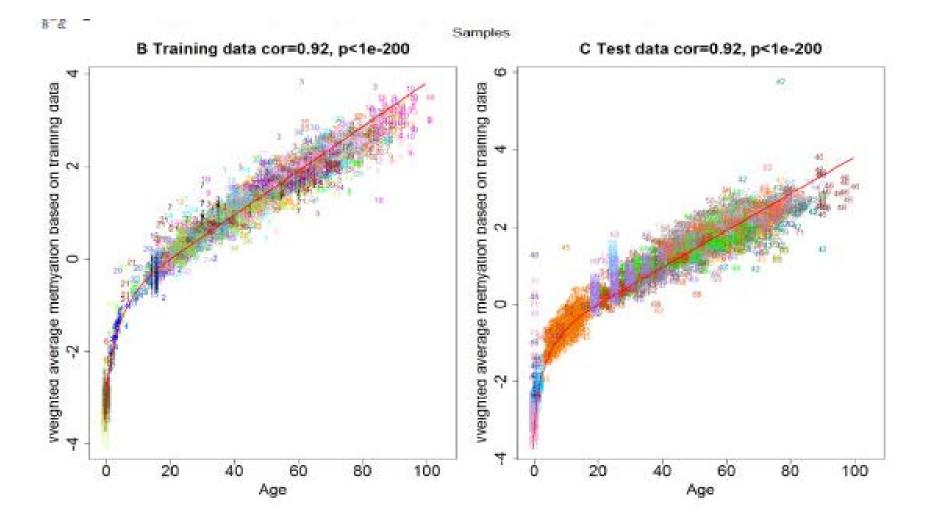
DNA methylation age of human tissues and cell types

Steve Horvath 🖾

 Genome Biology
 2013
 14:3156
 DOI: 10.1186/gb-2013-14-10-r115
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 10 June 2013
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DNA Methylation Age

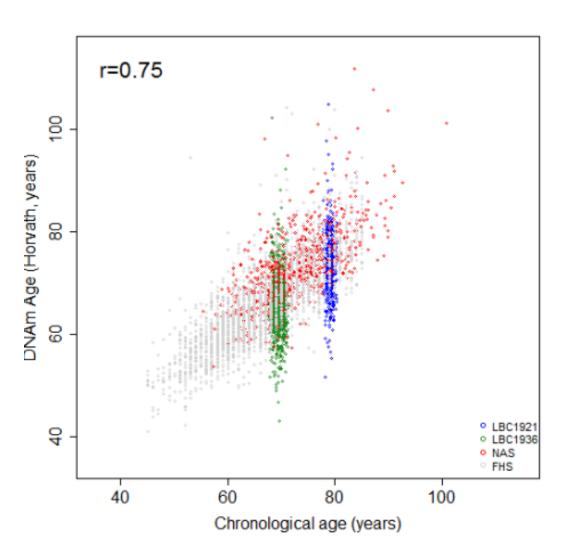


DNA methylation age

- Horvath demonstrates his measure:
 - Is applicable to a wide range of tissues
 - Works in chimpanzees
 - Stem cells have a DNA methylation age close to zero
 - Is negatively associated with number of mutations found in cancer cells

Age Acceleration

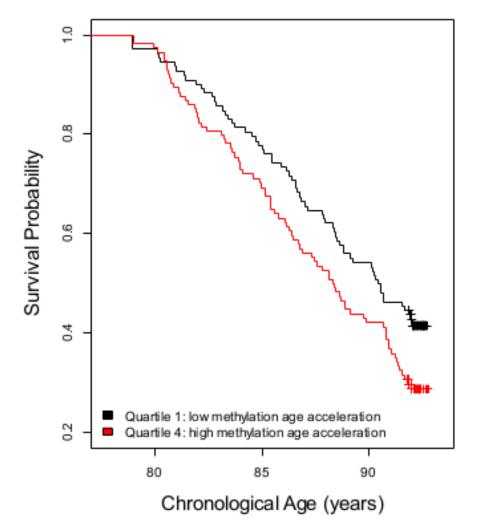
- There is variation in DNA methylation age for people with the same chronological age
- This is referred to as age acceleration
- Is this variation important?



Age Acceleration

- People with higher age acceleration have a higher rate of mortality than those with low age acceleration
- Effect still present after correcting for
 - Smoking
 - BMI
 - CVD
 - Removing people who died within five years of measurement
- Effect on mortality is independent of telomere length





Age Acceleration

- Has been associated with
 - Lung function
 - Grip strength
 - Cognition
 - Cardiovascular disease
 -

Age Acceleration is Heritable

