# PBG 200A Notes

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#### • SNP array genotyping

- based on DNA-DNA hybridization
- need to know where the variable sites are in an organism
  - \* multiple prob pairs for each variable nucleotide
- the individual who carries a C(A) will hybridize better to probes with G(T). Measuring differences in signal intensities.
- very precise, very cost-effective, on a per-individual, per-SNP basis
- only is useful in model organisms
- For \$600, get 1 million alleles in 1 individual.

#### • Bead array genotyping

- most common for nonmodel organisms,
- but have to have a reference genome and know the variable sites and alternative alleles at a site
- need the same info for nonmodel and model organisms, but for nonmodel organisms you need to make your own reagents.
- primers attach to reverse-primers which have their own barcode. Reverse primers attach to beads, so each type of bead corresponds to a single allele. Plate-reader fishes out the beads and immobilizes them in different locations on a plate. Image the plate for the different tags.
- advantage: flexibility.
  - \* \$.2 \$.5 per SNP per genotype (48-384 SNPs)
  - \* minimum experiment: approx 500 individuals, \$10,000-\$15,000.
  - \* high requirements for DNA quantity and quality

## • Sequenom MassArray genotyping

- have to know a reference sequence (something)
- need to know 100-200 variable sites
- do PCR for every locus
- no tags, no beads, directly reading the sequence with a mass spectrometer (must be able to determine each allele from each allele in all sites)
- main limitation is SNP compatibility
- -30-36 SNPs per plex.
- flexible: suitable for gradual experiments (feeds information a little bit at a time, can improve experimental design on the fly)
- main limitation is SNP compatibility
- \$.15-\$.25 per SNP per genotype
- requires less DNA and less quality

## • Genotyping by Sequence

- Do reads for k individuals in a single pool
- separate bioinformatically later on

- assign ancestry to each individual at each location
- Multiplexed shotgun genotyping
  - essentially getting unlimited info
  - assuming the individuals have high LD
  - if the density of variate sites is much less than the density of recombination break points.
  - this means we don't need high coverage we don't have complete genome
  - can be very cheap on the per individual basis can multiplex multiple individuals on a single Illumina lane.
  - very cheap \$20-\$40 per individual fo thousands of markers
  - disadvantages:
    - \* need to have fairly well-assembled reference genomes, and info on each parent,
    - \* high upfront costs (barcode adaotors need a single barcide for each individual). The same 100 or so barcodes can be reused. Scalaing up is easy.
  - Key point: not using genetic information to infer a genetic map. Using physical genome as reference.
- Bulked segregant analysis
  - Basic principle: identify differences in the frequencies of marker alleles between pools of individuals with different phenotypes
  - In the absence of population structure, this is useful (do we want most or least diverse group?)
  - can be done in lab or in natural populations
  - limit by scale of LD
  - Genetic designs for BSA
    - \* controlled lab cross
    - \* hitchhiking mapping (select for phenotypes, see which genotypes come with it)
    - \* introgression mapping
    - \* Natual poulaition
- Genotyping by sequencing (Gar paper): technology
  - a lot easier and simpler can't use it for genome mapping without a reference equilibrium
  - this was an excellent example of how to exploit the available technological resources
  - pseudogenomes (updated genome) mapping to a semi-related genome assume the genomes are close enough, and that they are more or less colinear. Map the reads to the semi-related genome - very high error rates, but fine for their purpose
  - masking polymorphisms throwing out over 90% of data, still end up with hundreds of variable site per Mb.
- Measuring gene expression: microarrays vs. RNA-seq
  - what genes are expressed where, when, and in what amount
  - determines the function of the genome
  - look at gene expression to determine the link between genotype and phenotype
  - Microarrays
    - \* immobilize probes on solid support
    - \* "analog" technology
    - \* cost-effective when good arrays already available,
    - \* but custom arrays can be made for any organism
    - \* has almost completely been displaced by...
  - RNA-seq
    - \* "digital" technology
    - \* can be done for any organism, but requires a reference for mapping reads
    - \* assuming the number of reads you observe in an RNA sample is proportional to expression

- \* don't need an organism-specifi reagent, but do need a reference genome,
- \* but can make a custom reference (which is a helluva lot easier than making a custom array)
- \* got cheaper fast, now even cheaper and easier
- Design and analysis of expression experiments
  - interested in differences between treatments/categories
    - \* brain/liver, grain/rice, male/female, etc.
    - \* must account for variation within treatment first!
  - biological variation within treatment
    - \* genetic differences, environmental conditions, etc.
  - technical variance
    - \* effect of experimental procedures, dissection, batch of arrays/dyes, reagents, ozone levels that day, etc.
    - \* Artyom knows a statistician who swears she can look at two data sets and determine which Illumina machine it came from..... wtf.
  - biological replicates necessary to detect differences
    - \* avoid confounding biological and technical variation
    - \* know the technical properties of your method
    - \* anticipate sources of biological and technical variance
- design of RNA-seq experiments
  - unlike microarrays, RNA-seq provides categorical (counts) data
  - in principle, can detect "differential expression" from a single replicate per treatment (conditioned on depth)
  - does not tell you whether you are looking at true biological variation
  - for quantitative experiments, need multiple biological replicates per treatment, just a with arrays
  - minimize environmental and technical variation
  - avoid confounding biological and technical variation
  - example of a bad design:
    - \* using different reagents, different Illumina machines completely confounding biological and technical variation
  - example of a good design:
    - \* use barcodes!
    - \* isolate RNA use the same batch of reagents for each library
    - \* can mix all six biological replicates because of barcodes
    - \* put the mix on six different lanes, minimizing technical variation
  - the importance of this depends on the level of variation you are measuring
  - may be looking at subtle quantitative difference
- problems with RNA-seq quantification
  - standard Illumina procedure includes many enzymatic steps with purification in between
  - losses and biases at every step
  - OK if looking for big differences; may be a problem if looking for subtle quantative variation
- Allele-specific expression analysis
  - Differences in gene expression between genotpes can be due either to mutation in that gene (cis-regulatory) or any changes in its upstream regulators (trans-regulatory)
  - For some applications, we don't care (health)
  - In evolutionary terms, however, it makes a big difference. How do we figure out where the divergence is based on cis- or transregulatory variation
  - F1 hybrids can be used to infer cisregulatory variation
  - By comparing two parents and their F1, can decompose gene expression divergence into cis- and trans0regulatory components