# Bioinformatics for transcriptional and genome variation

Kevin Murray
@kdmurray91
kevin@kdmurray.id.au

Borevitz Lab, ANU

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#### Overview

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- How we do RNAseq
- Experimental design: What we've done, how important it is
- RNAseq analysis pipelines
- DNA analysis pipelines



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- Wet lab: Mostly NEB/TruSeq kits; few attempts at custom library prep



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- Example experiment:
  - Arabdopsis (Col-0)
  - Three dynamic growth conditions
  - ▶ Before and after  $1000\mu\text{E}$  light treatment
  - 4 biological reps each group
  - Tissue harvested within 60 seconds of end of stress
  - TruSeq RNASeq kits, 12 samples/lane



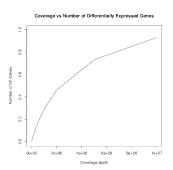
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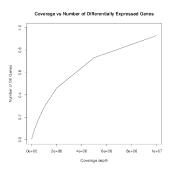


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 See Kliebenstein, (2012) FIPS: Exploring the Shallow End; Estimating Information Content in Transcriptomics Studies.



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- Open source, public:
- https://github.com/kdmurray91/RNAseqPipeline https://github.com/pedrocrisp/NGS-pipelines



## Sequence analysis pipeline



- ▶ fastqc
- scythe
- sickle
- subread/subjunc
- ▶ featurecounts
- edgeR
  - TMM normalisation
  - exactTest or glmFit
  - ► Also using limma's voom
- R scripts for post-analysis
  - ► GOseq
- Diagnostic plots highly recommended!

Robinson & Oshlack (2010); Liao et al. (2013a; 2013b); Robinson et al. (2013); Young et al. (2010)



## A Change of Pace

▶ Back to the slow, simple world of DNA for a moment...

- Genotyping-by-sequencing
  - ▶ Reference & de-novo analysis
  - Porting to NCI NF
  - Currently manual, working to automate
  - Processed > 5000 samples
- Reference-based genotype calling
  - Pipelines exist
  - Not used a lot, requires deeper coverage
  - See Norman's talk yesterday



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- e.g: gearing up to sequence 7500+ Eucalyptus, generating over 10 TB raw sequence data.
- k-mer analysis: analyse k-length words of sequence
  - ▶ Fast
  - Constant-memory (with khmer)
  - Scaleable (linear time w/ number of samples)
  - Parallelisable (within & across nodes)



#### **Thanks**

- Borevitz lab (Norman, Justin, Megan, Steve)
- Pogson lab (Pete Crisp)
- ► Genome Discovery Unit

► Slides at git.io/vfAof



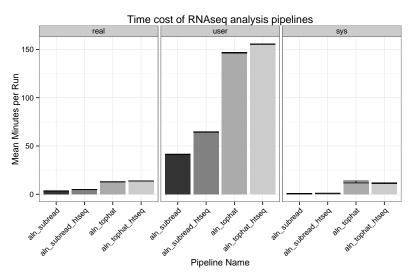
# Grab-bag of capabilities

- Confirm genotype using RNAseq reads
- Check technical reps are true



## Pipeline performance

▶ Faster than others by > 2 - 3x





#### MDS Plots save time!

▶ If your reps don't cluster, time to cry into beer.

