# Cancer phylogenetics using single-cell RNA-seq data

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J. M.

#### scRNA-seq is exciting!

#### Like scDNA-seq:

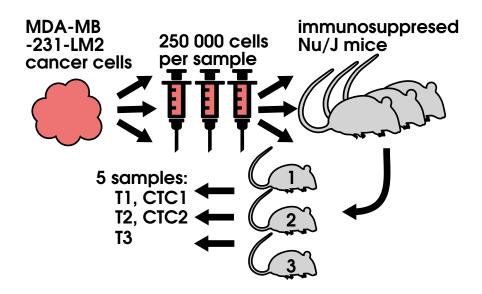
- guranteed single origin of DNA
- we can detect SNVs

#### But better!

- expression levels are commonly used
- expression levels are influenced by epigenetics
- we can detect CNVs

J. M. 2/2

#### **Experiment design**



J. M. 3/2

#### Data:

The expression levels for 43k genes varies greatly!

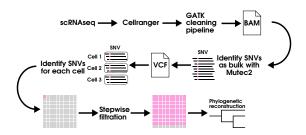
Sample	Cells	UMI	UMI/Cell	density	
TI	713	428k	600	0.66 %	
T2	2777	69k	25	0.052 %	
T3	810	652k	805	1 %	
CTC1	3108	129k	41	0.08 %	
CTC2	8275	28k	3	0.005 %	
total:	15683	1.3M	83	0.11 %	

Missing data are integral part of scRNAseq problem!

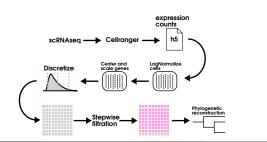
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#### Workflow

#### SNV:

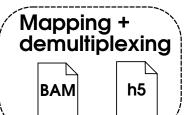


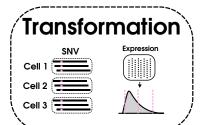
#### **Expression:**

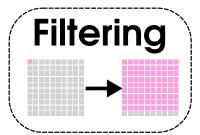


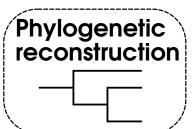
J. M. 5/2

## scRNAseq









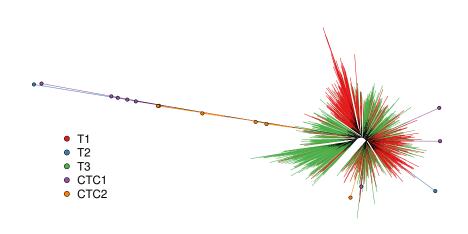
J. M. 6/2

#### **Trees**

- Expression and SNV
- Maximum Likelihood method using IQtree
- Progressive filtering to remove missing data
- filtered datasets: 20%, 50% and 90% data density
- running time for IQtree on the 20% expr: 6 hours

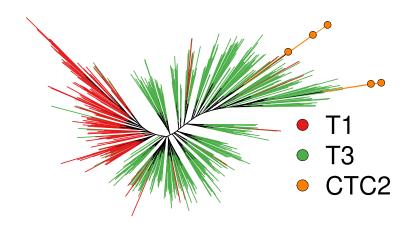
J. M. 7/2

## **Expression 20% density**



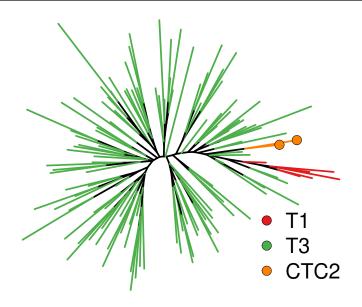
J. M. 8/2

## **Expression** 50% density



J. M. 9

## **Expression** 90% density



J. M. 10

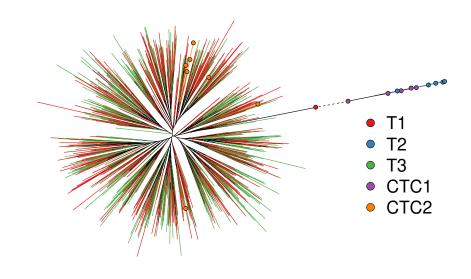
## SNV 20% density





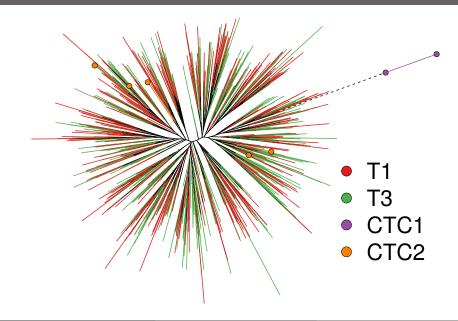
J. M. 11/21

#### SNV 20% density



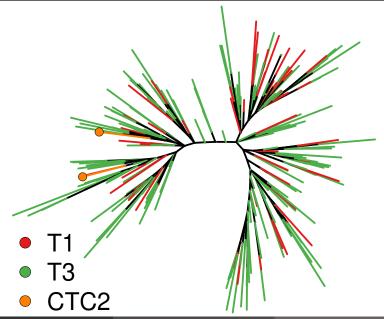
J. M. 12

## SNV 50% density



J. M. 13,

## SNV 90% density



J. M. 14/

#### Conclusion

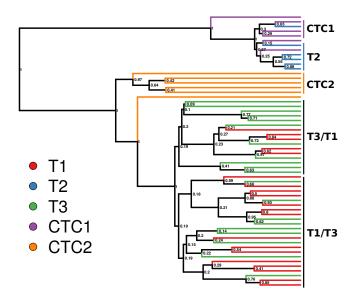
- Expression works well!
- SNV seems to fail.
- Statistical support for these trees is deceivingly high
- Bayesian analysis required, but datasets are too big or missing samples

#### New filtering approach

- Data reduced to 58 sequences:
  - 20 sequences for T1 and T3
  - 6 sequences for T2, CTC1 and CTC2
- dataset filtered to:
  - 10% density (full dataset)
  - 50%
  - 90%
- Bayesian analysis using BEAST

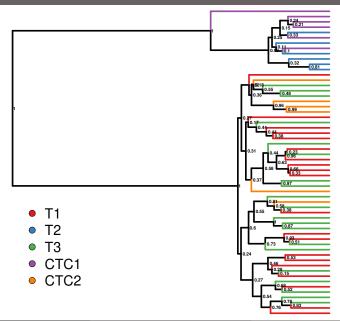
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#### SNV 10% data density tree from Bayesian analysis



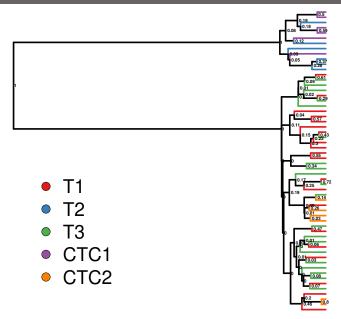
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#### SNV 50% data density tree from Bayesian analysis



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### SNV 90% data density tree from Bayesian analysis



J. M. 19/

#### Conclusion

- scRNA-seq does contain phylogenetic signal!
- decent phylogeny can be constructed from both expression and SNVs
- a lot of space for improvement: scRNA-seq callers and error models

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#### Acknowledgment







#### MINISTRY OF BUSINESS, INNOVATION & EMPLOYMENT HĪKINA WHAKATUTUKI

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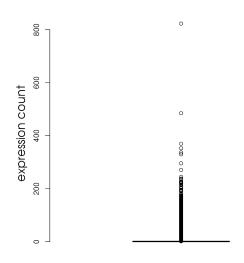


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## **Supplementary materials**

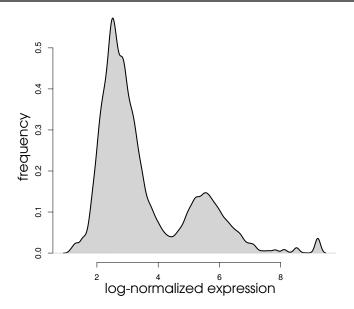
J. M. 1/

## **Boxplot of expression counts**



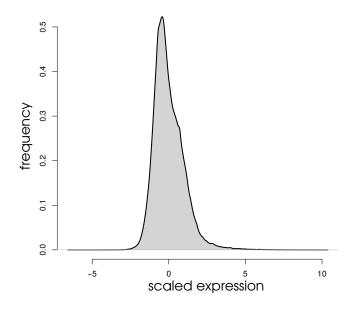
J. M. 2,

## **Log-normalization**



J. M. 3/

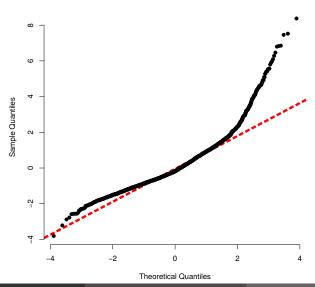
## Scaling: $\mu = 0$ , $\sigma = 1$

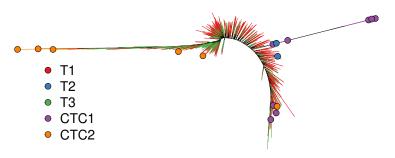


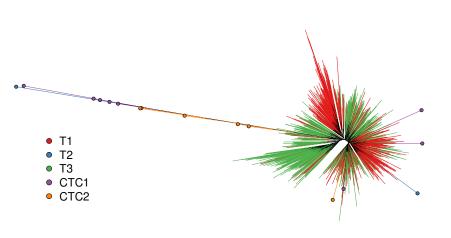
J. M. 4

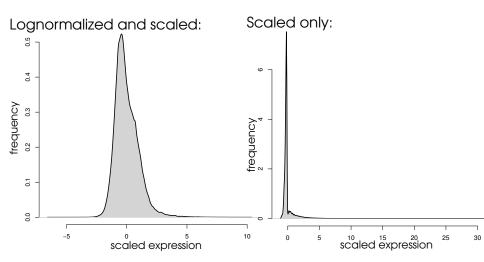
## Normality test









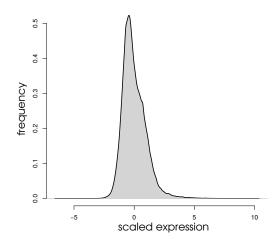


#### **Discretization**

**Problem:** Divide expression into groups.

Possible solutions:

- Centering around:
  - 0
  - mean
  - modus
- Intervals:
  - symetric
  - quantiles
  - HDI



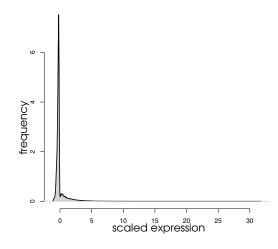
J. M. 9/14

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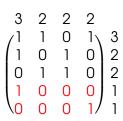
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#### Filtering data

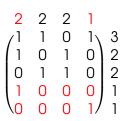
- Calculate sums of columns and rows (colsum, rowsum)
- Find smallest colsum/rowsum
- Substract the least represented columns/rows from the rowsum/colsum
- Remove the columns/rows from the matrix



J. M. 11/14

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J. M. 12/14

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#### Bootstrap and posterior scores:

		Matrix density 0.2 0.5 0.9		
Bootstrap	Expression SNV		77.36 70.99	
Posterior	SNV	53	54	24

#### Statistical significance:

- Bootstrap > 70
- Posterior > 95

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