Distance Measures for Tumor Evolutionary Trees

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Clonal Theory (Nowell 1976)



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Clonal Tree Inference



Clonal Tree Inference



How do we compare two clonal trees?



Introduction

- Many Trees
- Uses for Distance Measures
- Existing Distance Measures

2 Methods









Multiple Inference Methods



Inter-Patient Comparison



Intra-Patient Comparison







Uses

Comparing/evaluating inferred trees



Uses

- Ocomparing/evaluating inferred trees
- Olustering trees



Uses

- Comparing/evaluating inferred trees
- Olustering trees
- S Inference/consensus methods (e.g., Govek et al. 2018)

Existing Distance Measures



Phylogenetic trees

- Robinson-Foulds distance (Robinson & Foulds 1981)
- Quartet distance (Estabrook et al. 1985)
- Triplet distance (Critchlow et al. 1996)

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Clonal trees

Existing Distance Measures



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Clonal trees

- MLTED (Karpov et al. 2018)
- A-D distance (Govek et al. 2018)
- Rearrangement distance (Bernardini et al. 2019)





Methods

- Definitions
- CASet
- DISC

Results





Definition

A *clonal tree* is a multi-labeled tree with unique labels.

Clonal trees:



Definition

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Definition

An *m*-clonal tree is a clonal tree with labels $1, \ldots, m$.

Clonal trees:

7-clonal trees:



















Ancestral and Common Ancestor Sets



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Ancestral and Common Ancestor Sets



Common Ancestor Set

Given a clonal tree T_k and two mutations i, j,

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C_k(i,j) = A_k(i) \cap A_k(j).
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$$\mathsf{Jacc}(A,B) = 1 - \frac{|A \cap B|}{|A \cup B|}, \qquad \mathsf{Jacc}(\emptyset, \emptyset) = 0$$

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Theorem (e.g., Gilbert 1972)

Jaccard distance is a metric on sets.



Average over all values \rightarrow CASet(T_1, T_2) = 0.39



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CASet Distance (Common Ancestor Set)

Given two *m*-clonal trees T_k , T_ℓ ,

$$\mathsf{CASet}(\mathit{T}_k, \mathit{T}_\ell) = \frac{1}{\binom{m}{2}} \sum_{\{i,j\} \subseteq [m]} \mathsf{Jacc}(\mathit{C}_k(i,j), \mathit{C}_\ell(i,j)).$$





Distinctly Inherited Sets



Distinctly Inherited Sets



Distinctly Inherited Sets

Given a clonal tree T_k and two mutations i, j,

 $D_k(i,j) = A_k(i) \setminus A_k(j).$





DISC Distance (Distinctly Inherited Set Comparison)

Given two *m*-clonal trees T_k, T_ℓ ,

$$\mathsf{DISC}(T_k, T_\ell) = \frac{1}{m(m-1)} \sum_{\substack{(i,j) \in [m]^2 \\ i \neq j}} \mathsf{Jacc}(D_k(i,j), D_\ell(i,j)).$$

Metric Properties





$$= \frac{1}{\binom{m}{2}} \sum_{\{i,j\}\subseteq [m]} \operatorname{Jacc}(C_k(i,j), C_\ell(i,j))$$

$\mathsf{DISC}(T_k, T_\ell)$



$$= \frac{1}{m(m-1)} \sum_{\substack{(i,j) \in [m]^2 \\ i \neq j}} \operatorname{Jacc}(D_k(i,j), D_\ell(i,j))$$

Metric Properties



Theorem

CASet and DISC are metrics on m-clonal trees.

Different Label Sets: Union



Different Label Sets: Union



Different Label Sets: Intersection



Different Label Sets: Intersection



Overview of Distance Measures



Outline



Simulated Data

OncoLib (El-Kebir Group 2018)



Simulated Data













Intra-Family Structure



Family E

- Triple negative breast cancer (Wang et al. 2014)
 - Single-cell seq. at $72 \times$ coverage
 - Bulk deep seq. at $118,743 \times$ coverage
- Breast cancer xenograft (Eirew et al. 2015)
 - Whole-genome seq. at 35-72× coverage
 - MiSeq targeted deep seq.



TNBC Tree Inference Analysis



DiNardo et al. (Carleton College)

TNBC Tree Inference Analysis



Xenoengraftment Tree Inference Analysis



DiNardo et al. (Carleton College)

May 3, 2019 29 / 31

Xenoengraftment Tree Inference Analysis



LICHeE inference

- Oistance measures are important for tumor tree analysis
- We introduced two novel distance metrics, CASet and DISC
- OASet clusters trees more clearly than existing measures
- OCASet and DISC have high resolution on simulated and real data



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Availability

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Implementations of CASet and DISC are available at
https://bitbucket.org/oesperlab/stereodist.
Preprint available at
https://www.biorxiv.org/.
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