

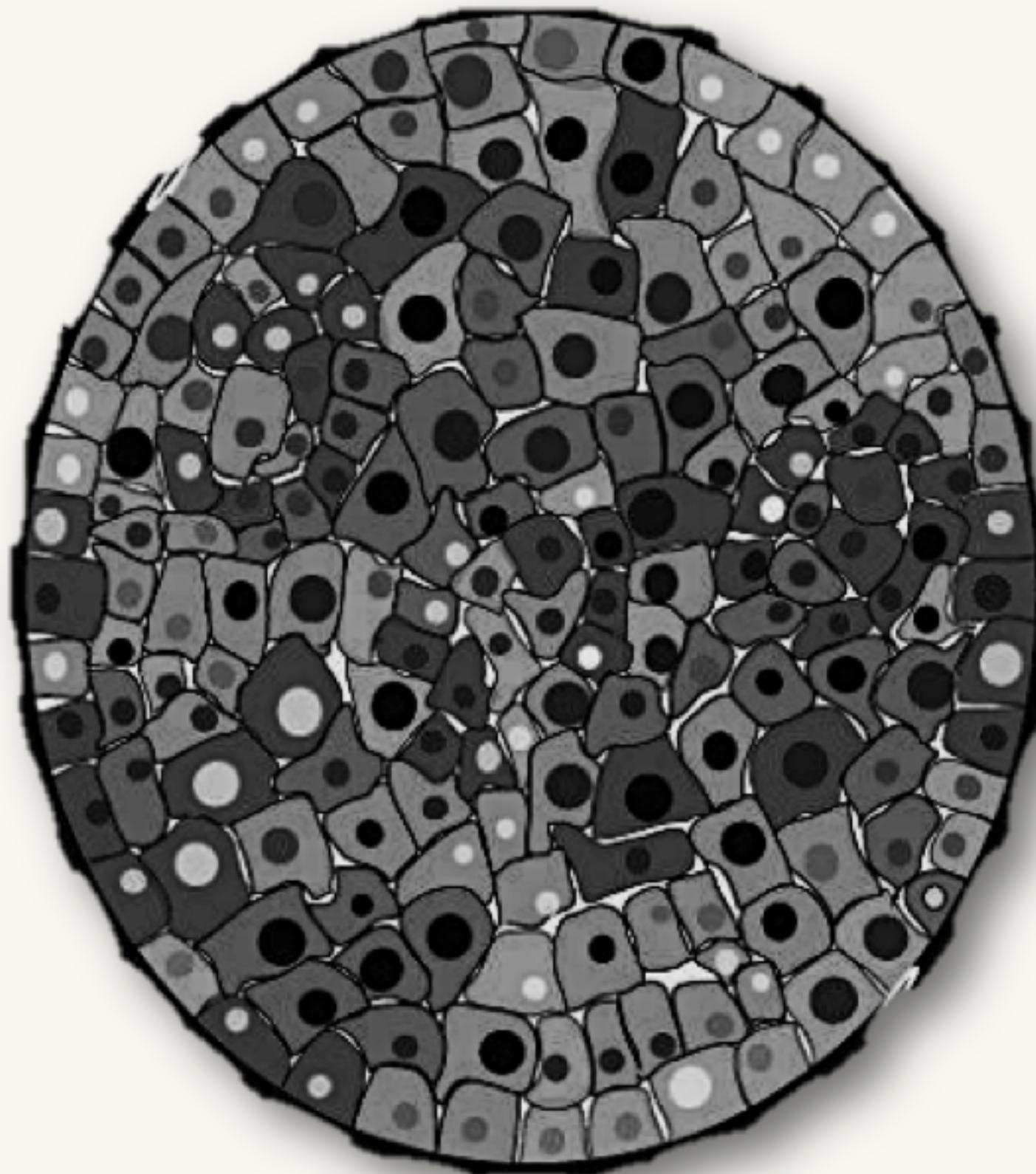
Cancer Phylogeny Inference Using Multi-Sample Somatic Variants

Victoria Popic

PhD Student in *Batzoglou Lab*
Stanford University



Intra-Tumor Heterogeneity

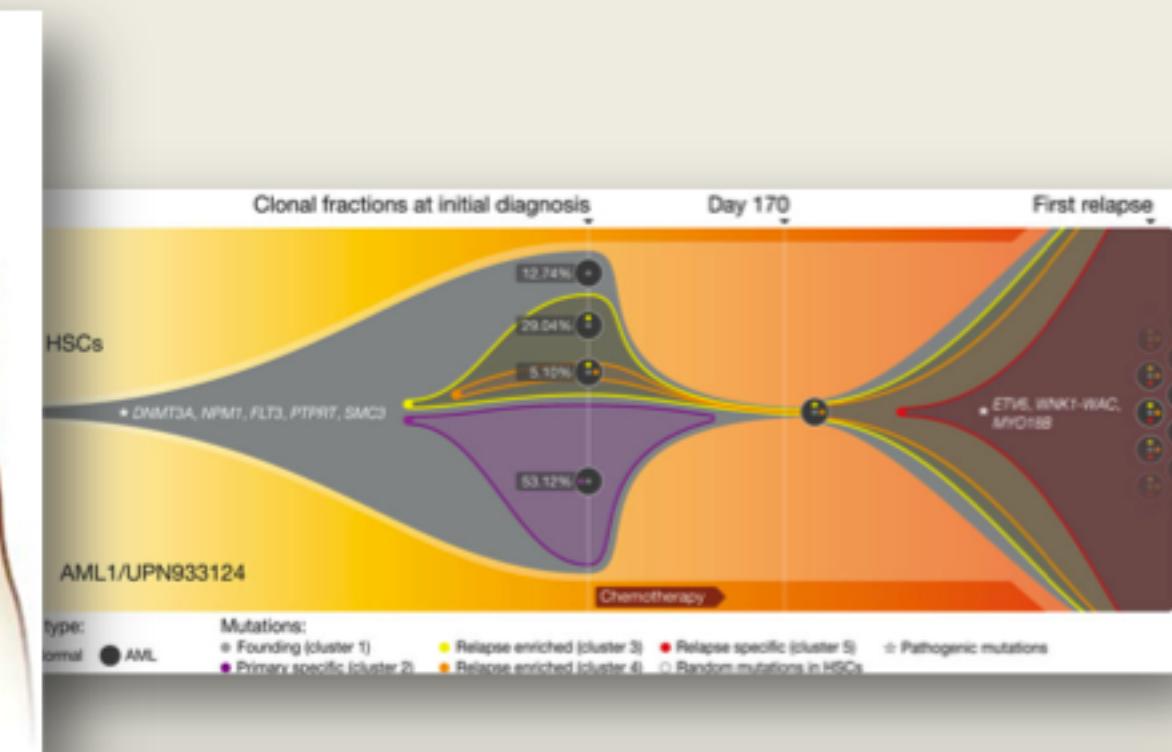
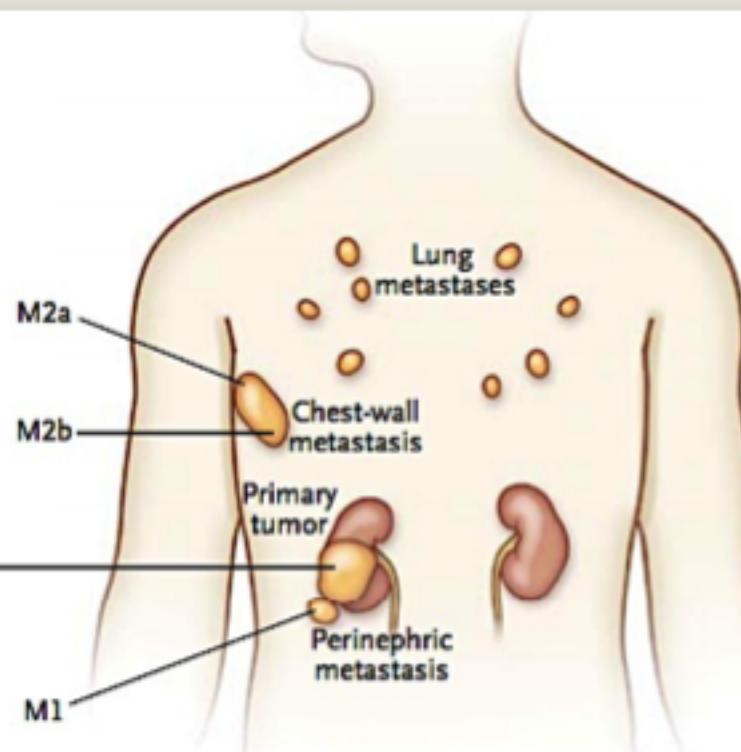
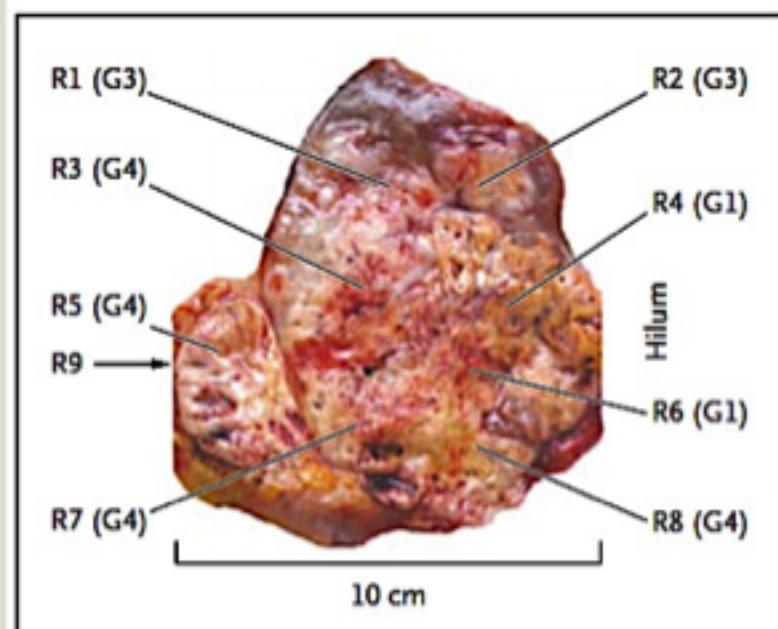


Intra-Tumor Heterogeneity



Multi-Sample Sequencing Studies

Biopsy Sites

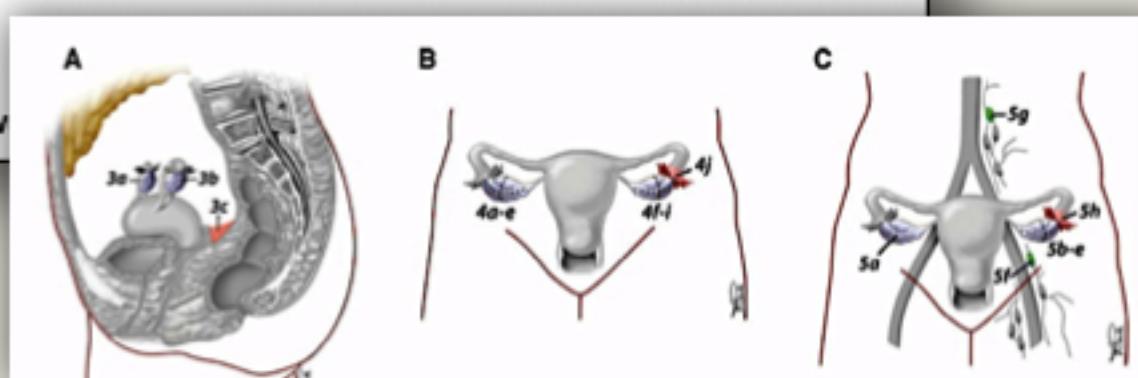


Intratumor Heterogeneity and Branched Evolution Revealed by Multiregion Sequencing

Marco Gerlinger, M.D., Andrew J. Rowan, B.Sc., Stuart Horswell, M.Math., James Larkin, M.D., Ph.D., David

Genomic architecture and evolution of clear cell renal cell carcinomas defined by multiregion sequencing

Marco Gerlinger, Stuart Horsw



Clonal evolution in relapsed acute myeloid leukaemia revealed by whole-genome sequencing

Li Ding, Timothy J. Ley, David E. Larson, Christopher A. Miller, Daniel C. Koboldt, John S.

Genome evolution during progression to breast cancer

Daniel E. Newburger^{1,6}, Dorna Kashef-Haghghi^{2,6}, Ziming Weng^{3,6}, Raheleh Salari², Robert T. Sweeney³, Alayne L. Brunner³, Shirley X. Zhu³, Xiangqian Guo³, Sushama Varma³, Megan L. Troxell⁴, Robert B. West^{3,7}, Serafim Batzoglou^{2,7} and Arend Sidow^{3,5,7}

Distinct evolutionary trajectories of primary high-grade serous ovarian cancers revealed through spatial mutational profiling.

Bashashati A¹, Ha G, Tone A, Ding J, Prentice LM, Roth A, Rosner J, Shumansky K, Kalloger S, Senz J, Yang W, McConechy M, Melnyk N, Anglesio M, Luk MT, Tse K, Zeng T, Moore R, Zhao Y, Marra MA, Gilks B, Yip S, Huntsman DG, McAlpine JN, Shah SP.

LICHeE*: Fast and Scalable Inference of Multi-Sample Cancer Lineages

Popic V, Salari R, Hajirasouliha I, Kashef-Haghghi D, West RB, Batzoglou S.
ASHG, 2014
Genome Biology, 2015



Raheleh Salari



Rob West



Arend Sidow

Input Data

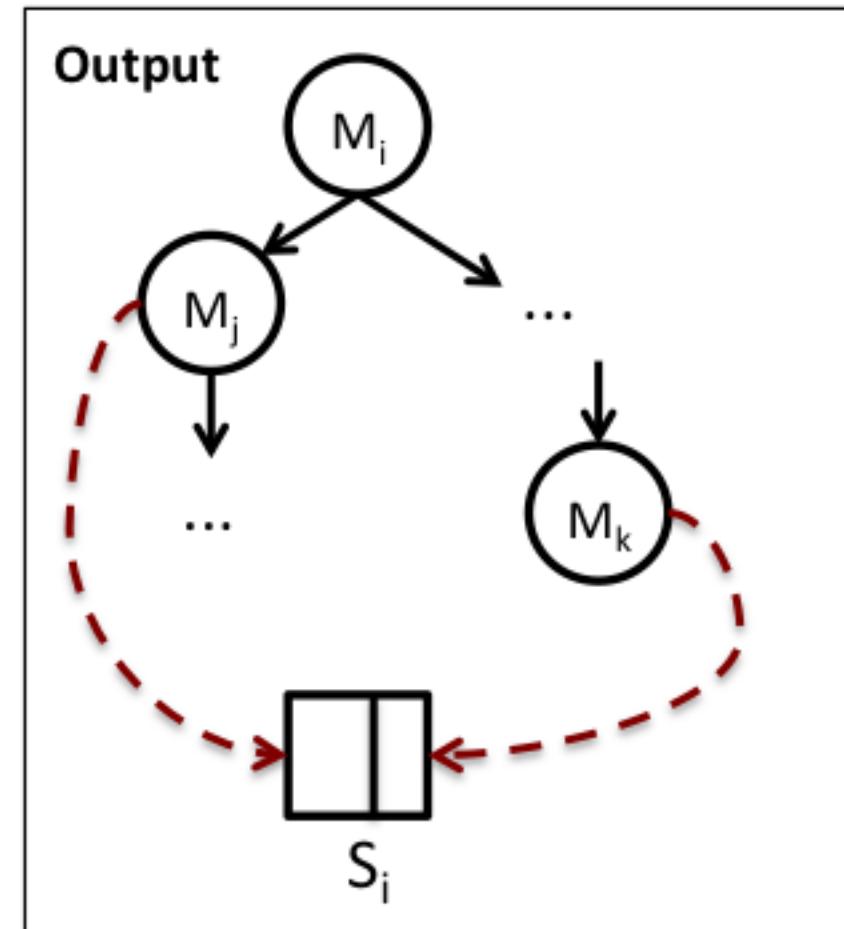
Single Nucleotide Variants (SNVs)

	#chr	position	description
M_1	1	184306474	A/G HMCN1
M_2	1	18534005	C/A IGSF21
M_3	1	110456920	G/A UBL4B
...			
M_N	10	26503064	C/G MYO3A

Variant allele frequencies (VAFs) per sample

Normal	S_1	S_2	S_3	...	S_M
0.0	0.1	0.2	0.25	0.15	
0.0	0.1	0.25	0.2	0.1	
0.0	0.4	0.4	0.45	0.45	
0.0	0.4	0.0	0.0	0.24	

Note: In general, the method can handle any type of variant given its cell prevalence (CP) values in each sample



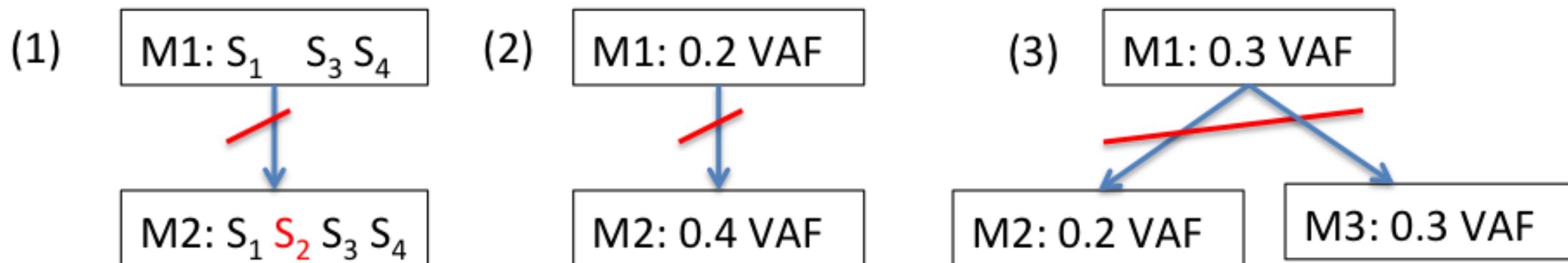
Perfect Phylogeny Model Assumption

Mutations **do not recur independently** in different cells
⇒ cells sharing the same mutation must have inherited it
from a **common ancestral cell**

Perfect Phylogeny Model: Constraints

Three SNV Ordering Constraints:

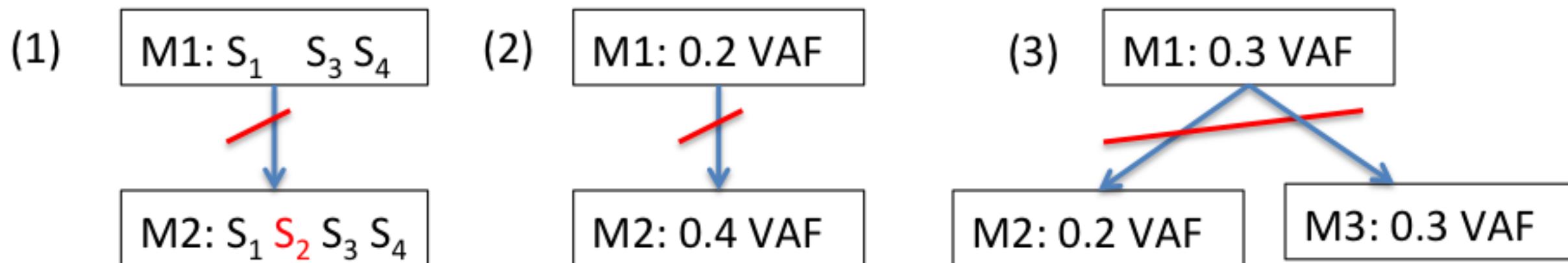
1. a mutation present in a given set of samples cannot be a successor of a mutation present in a smaller subset of these samples
2. a mutation cannot have a VAF higher than that of its predecessor mutation (except due to CNVs)
3. the sum of the VAFs of mutations disjointly present in distinct subclones cannot exceed the VAF of a common predecessor mutation present in these subclones



Perfect Phylogeny Model: Constraints

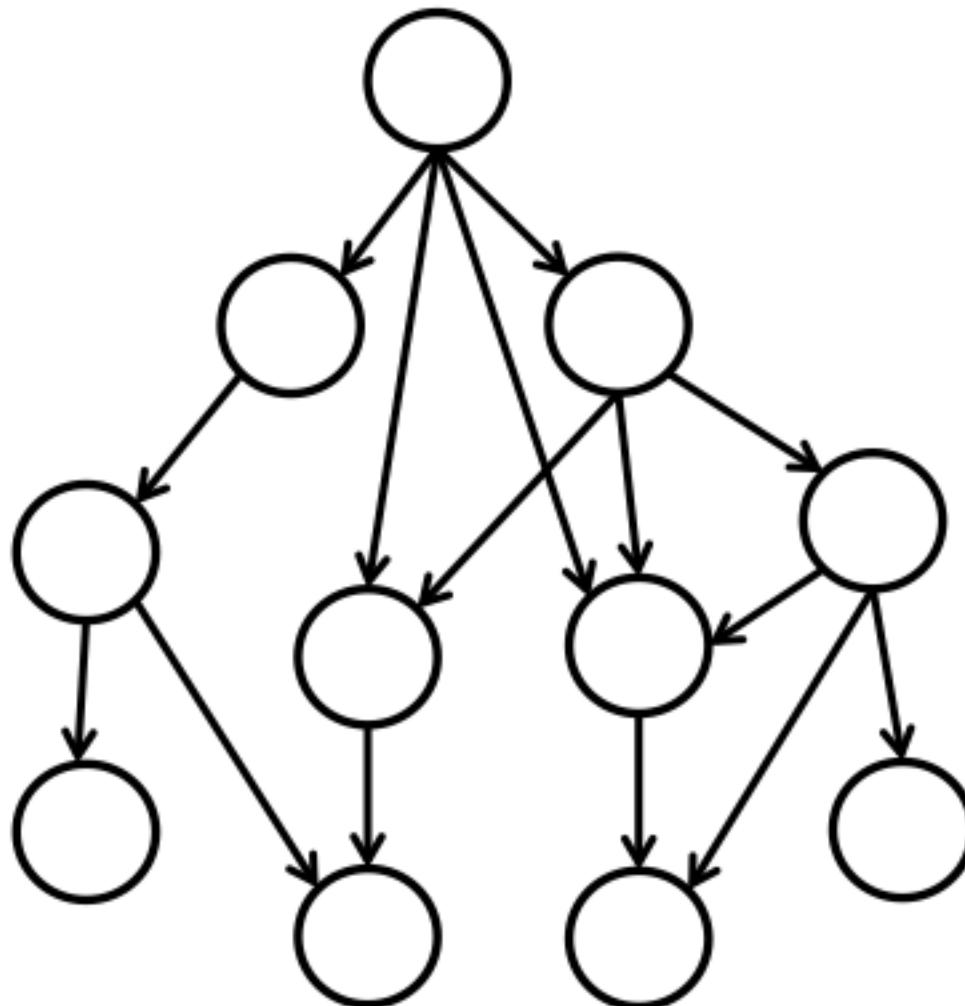
Three SNV Ordering Constraints:

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Goal: find all lineage trees that satisfy the above three constraints

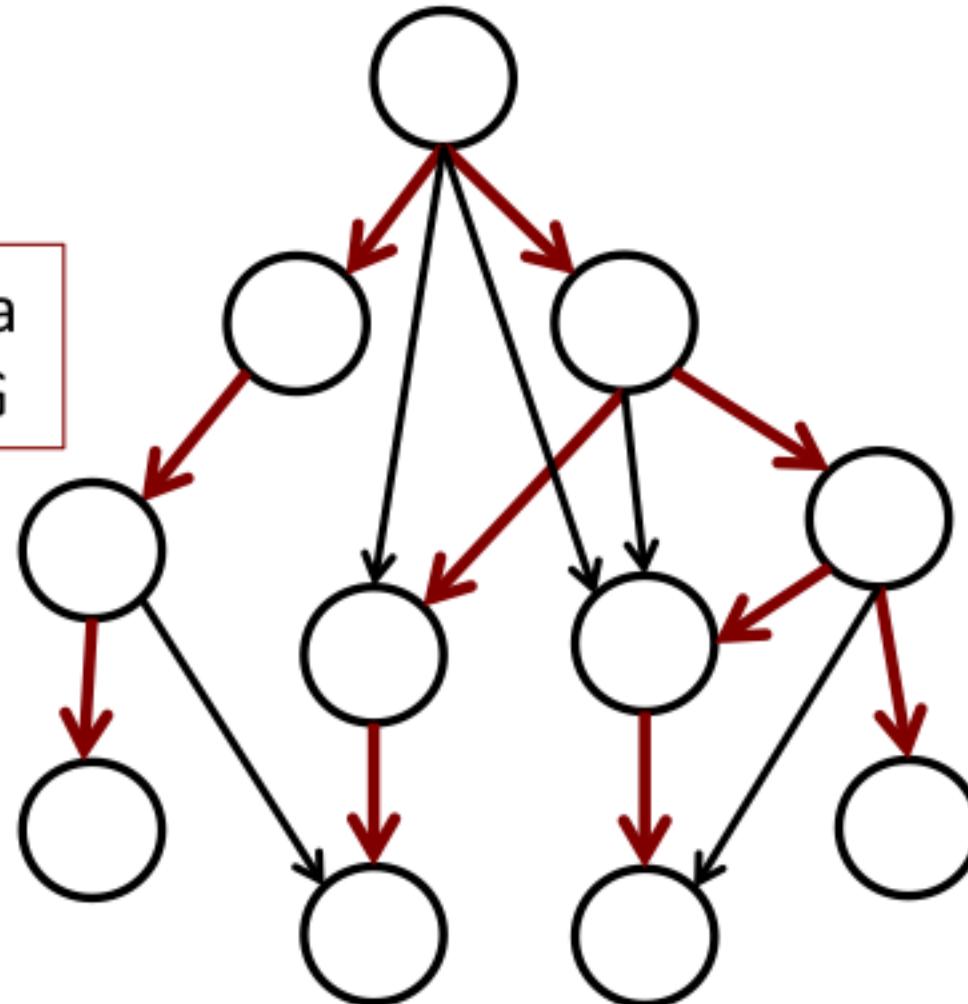
LICHeE's Problem Formulation



DAG encoding all pairwise
valid precedence
relationships –
*evolutionary constraint
network*

LICHeE's Problem Formulation

True lineage tree will be a spanning tree of this DAG



DAG encoding all pairwise valid precedence relationships – *evolutionary constraint network*

→ search for all lineage trees that satisfy constraint (3)

LICHeE: Method Overview



Given: SSNV multi-sample VAFs

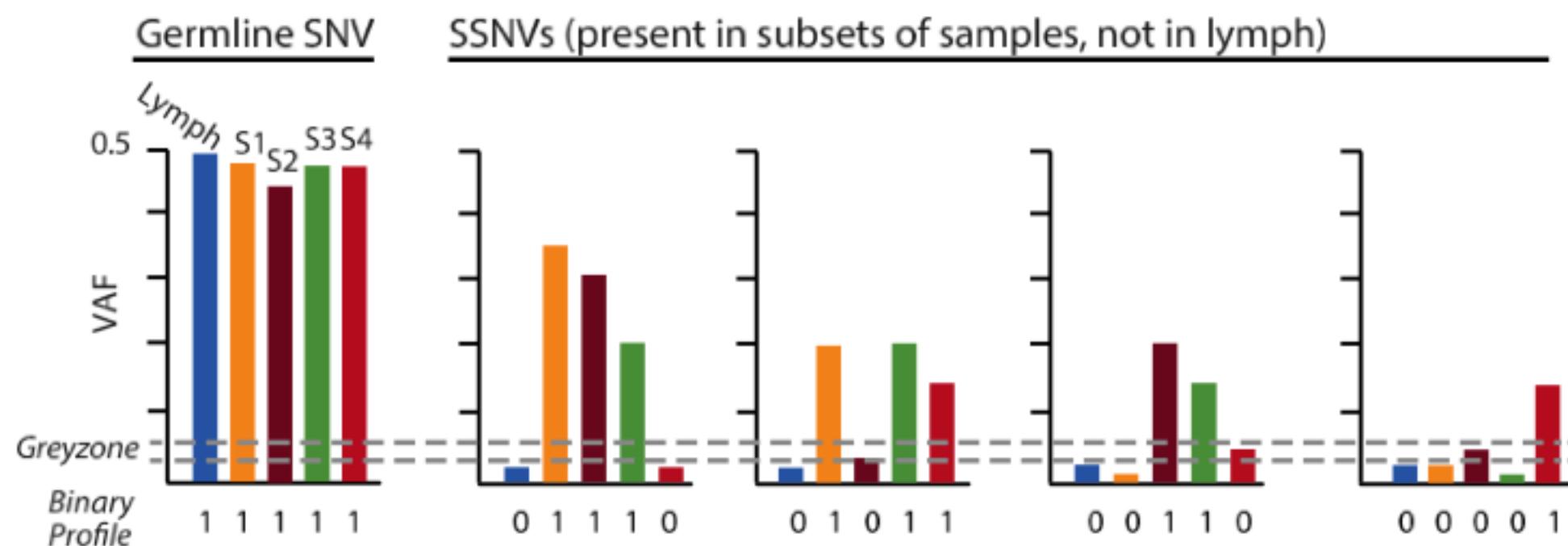
Algorithm steps:

1. Grouping and clustering SSNVs
2. Evolutionary Constraint Network Construction
3. Lineage Tree Search and Ranking

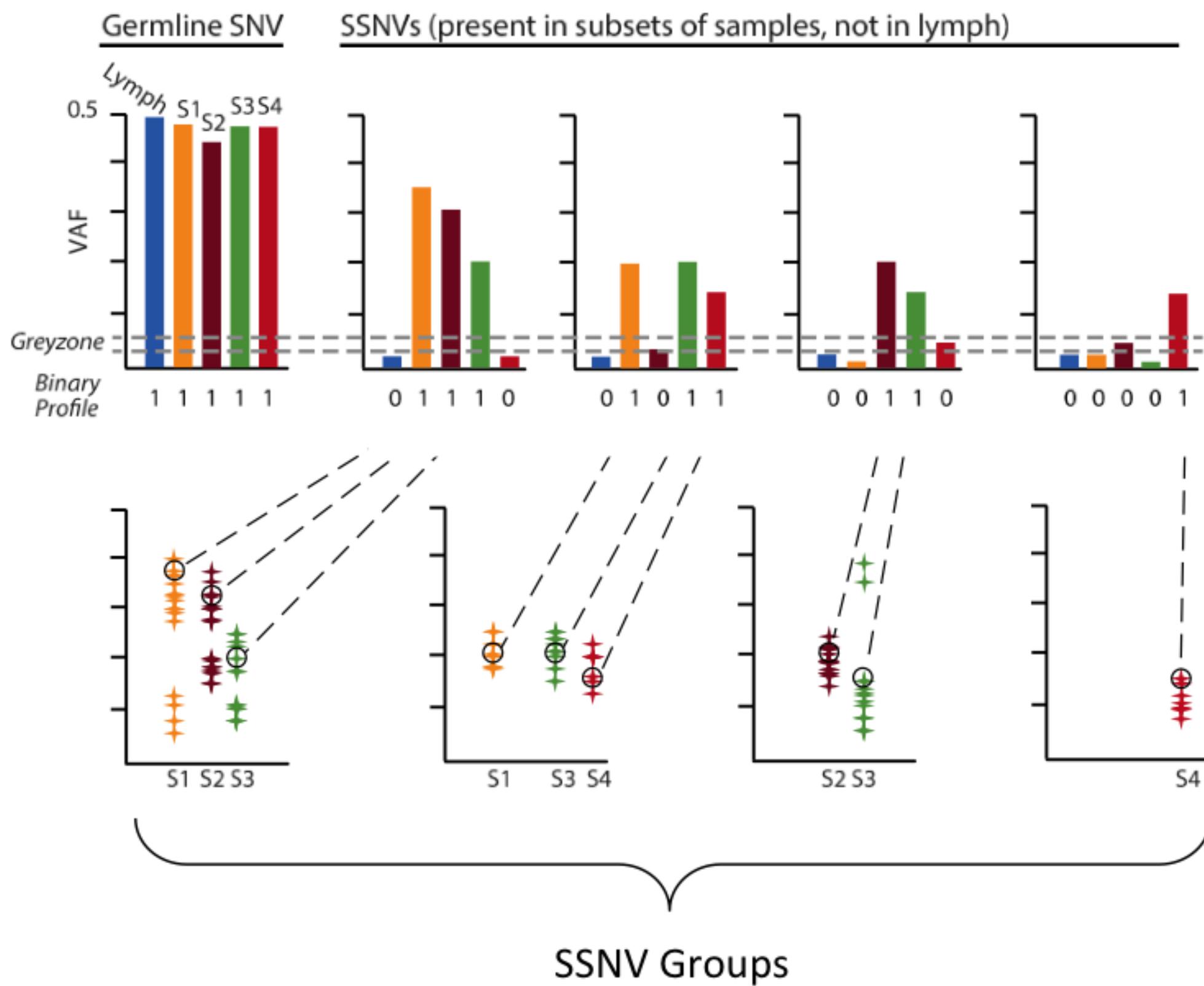
1. Grouping and clustering SSNVs

- presence patterns across samples
- VAF similarity

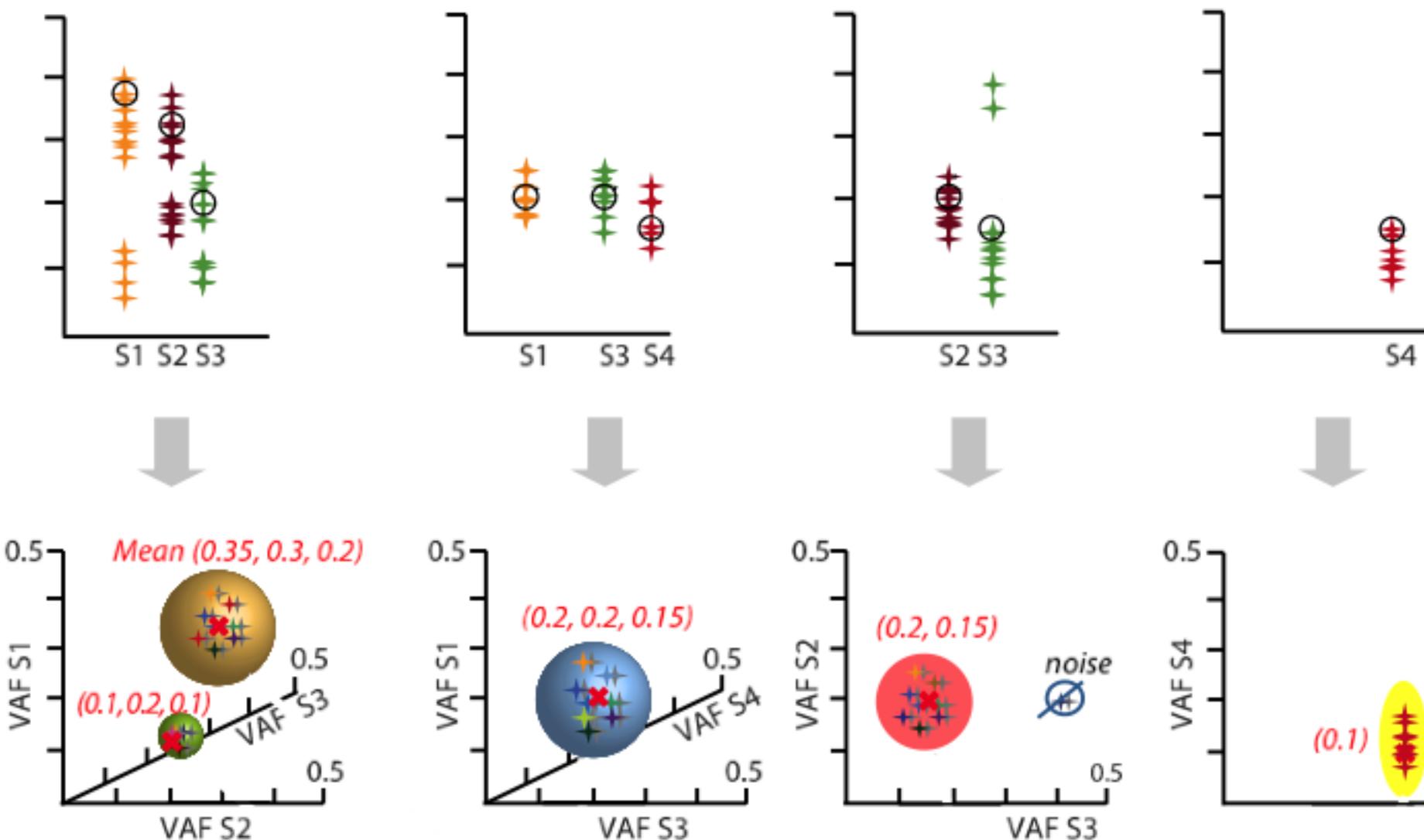
Presence Patterns Across Samples



Presence Patterns Across Samples



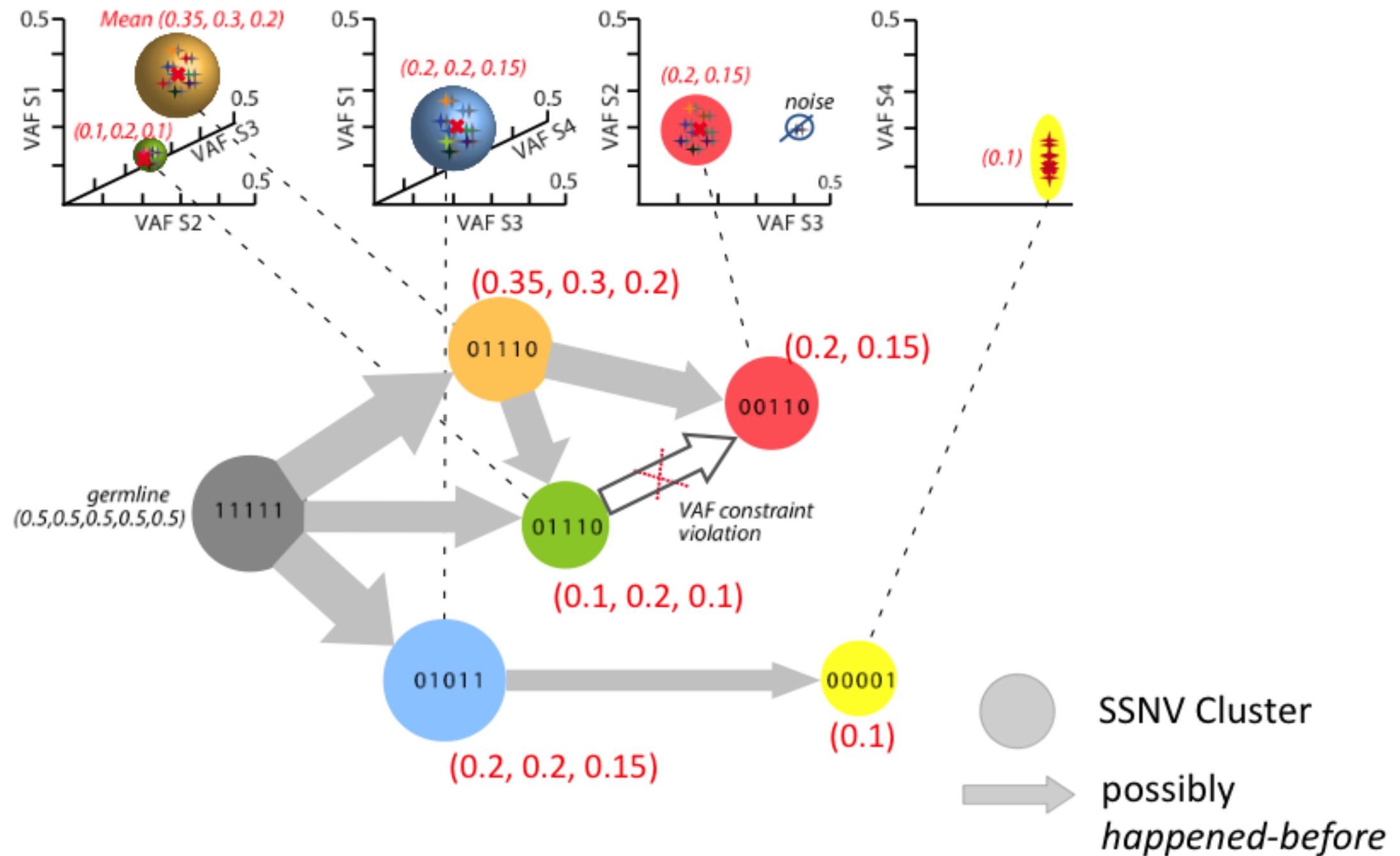
VAF-Based Clustering



2. Evolutionary Constraint Network Construction

- encodes whether a given cluster of SSNVs could have preceded another
- valid lineage trees are embedded in this network

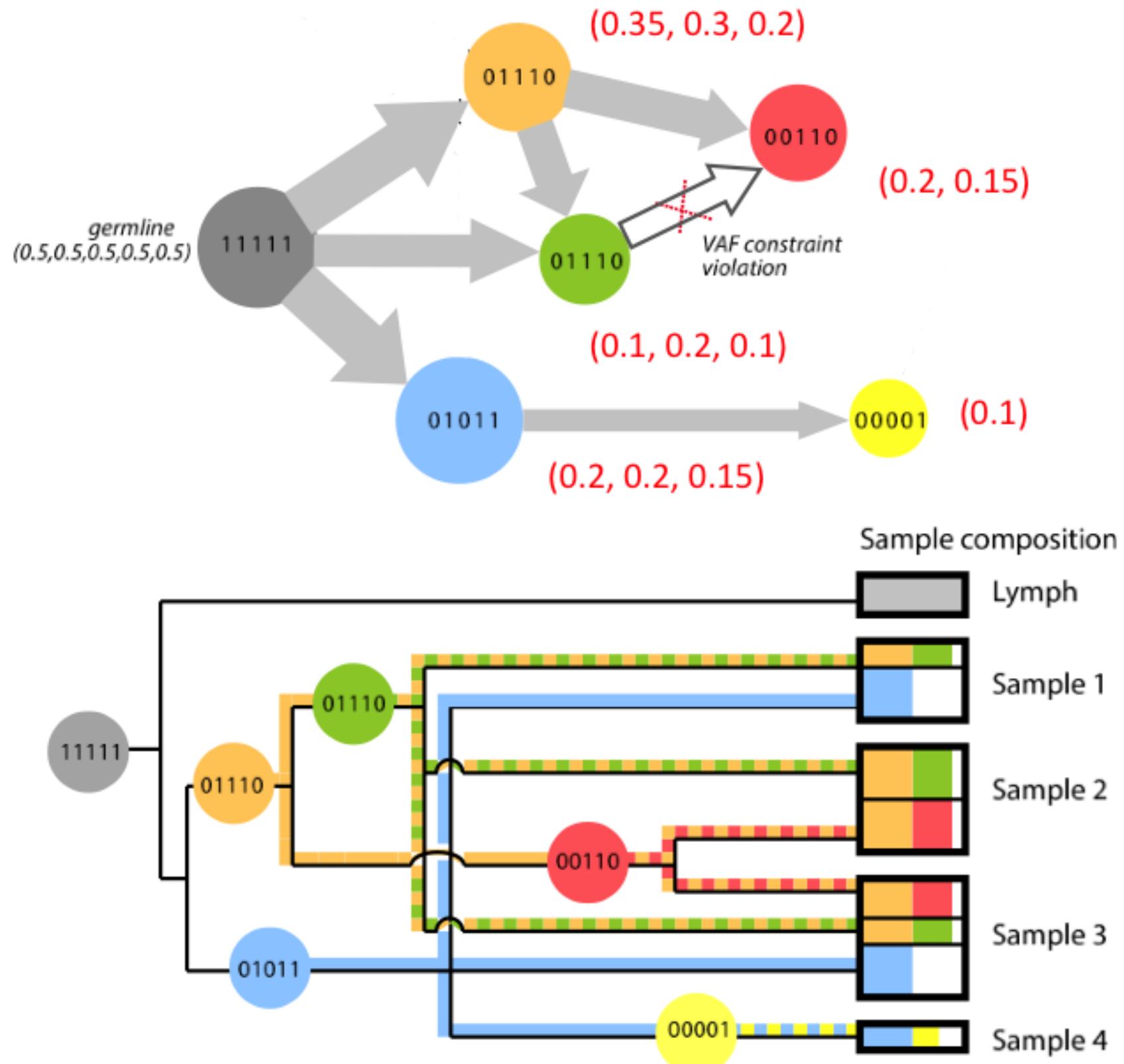
Evolutionary Constraint Network



3. Lineage Tree Search and Ranking

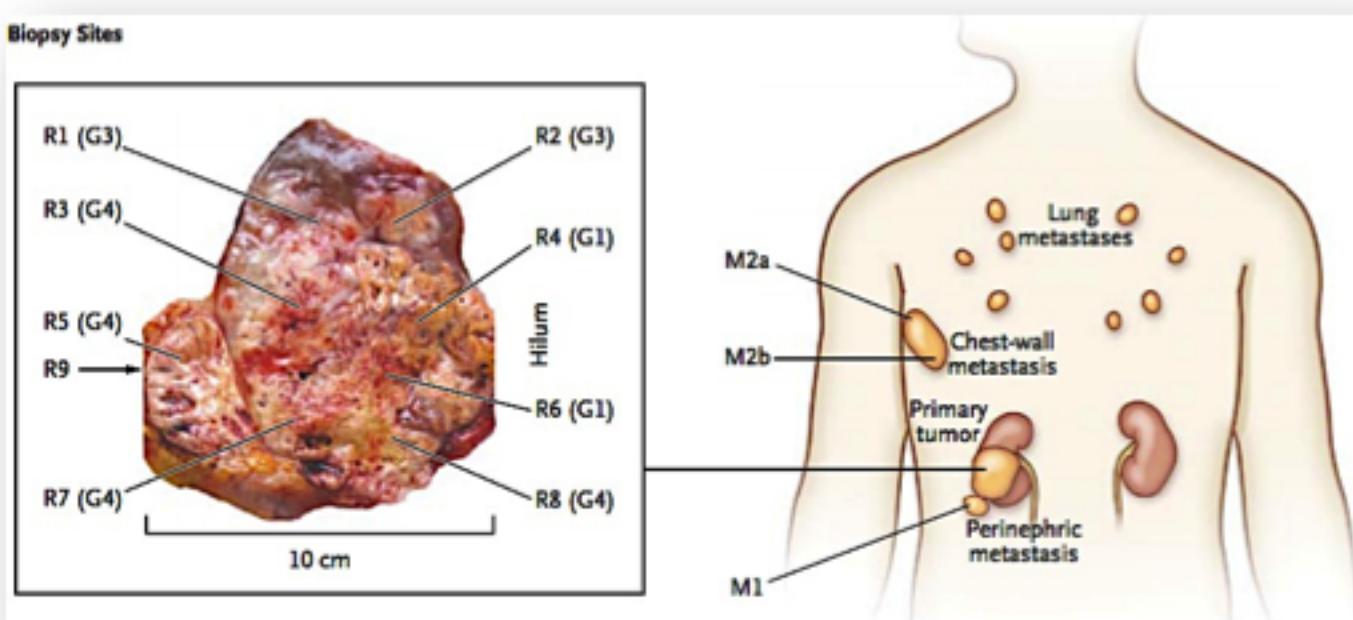
- search for spanning trees satisfying VAF constraints
within an error margin (extension of Gabow and Myers'78)
- top tree minimizes the squared deviation from the
cluster centroids

Lineage Tree Search

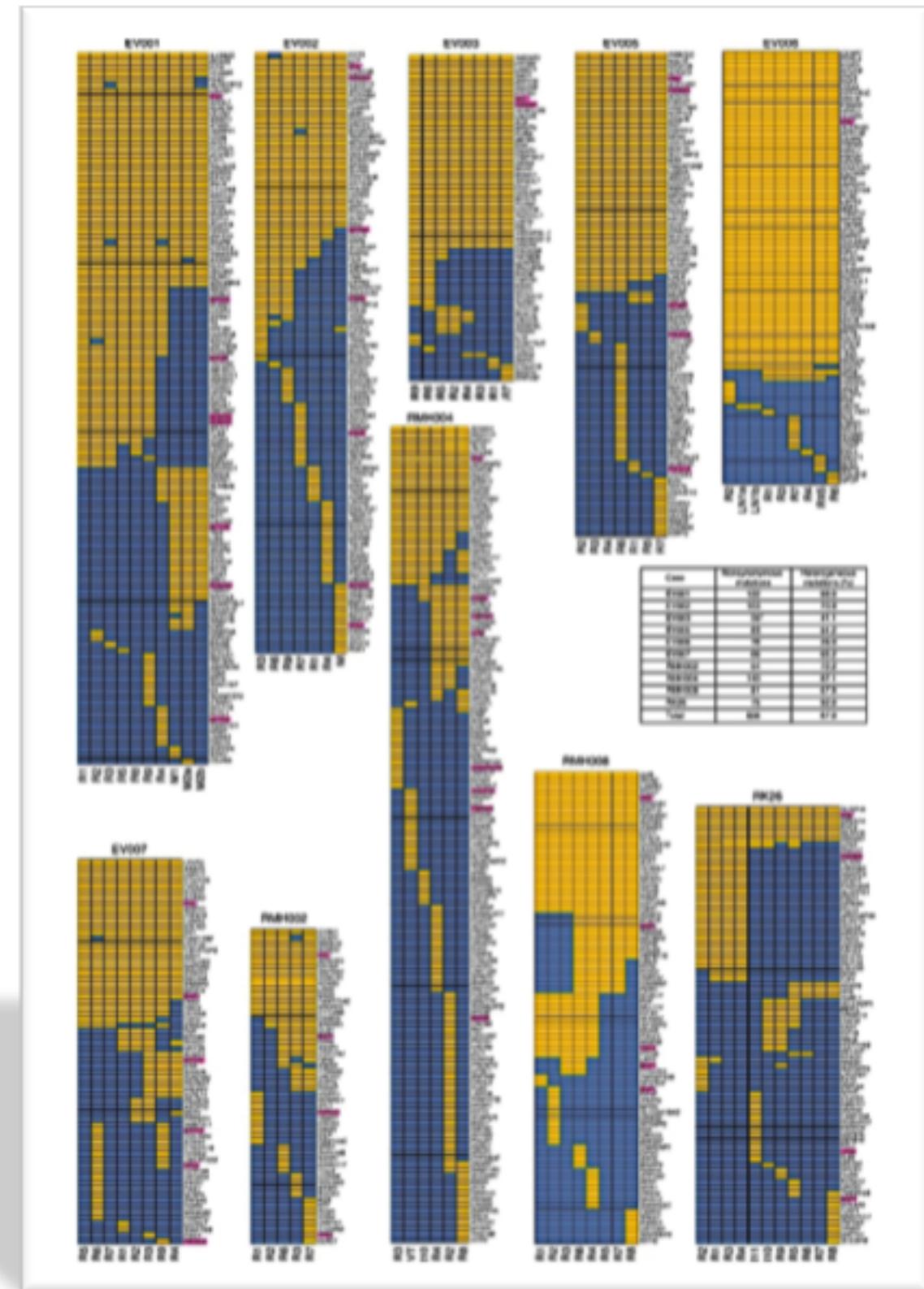


RESULTS

ccRCC Study by Gerlinger *et. al* (2014)

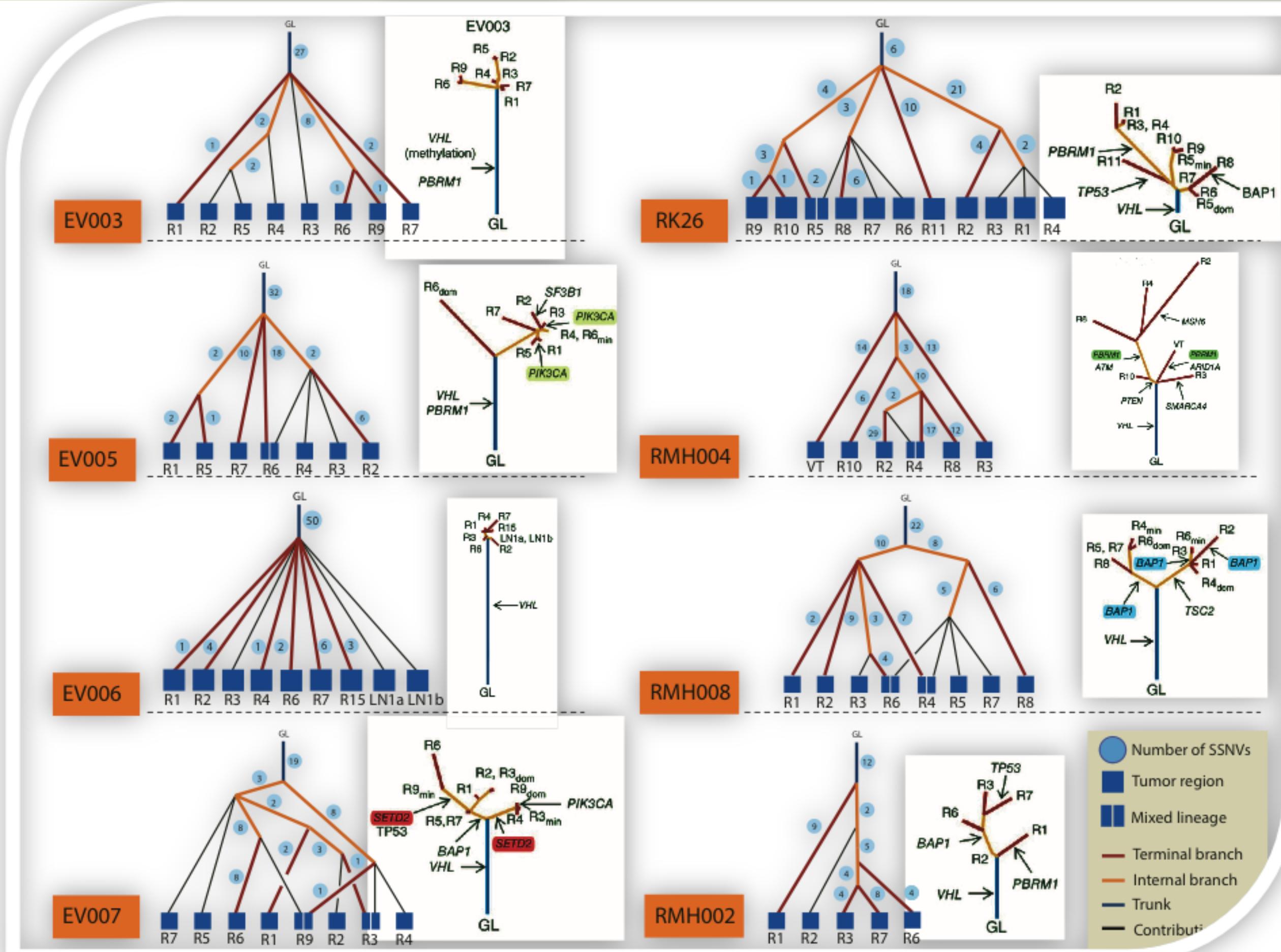


8 patients, 587 SNVs

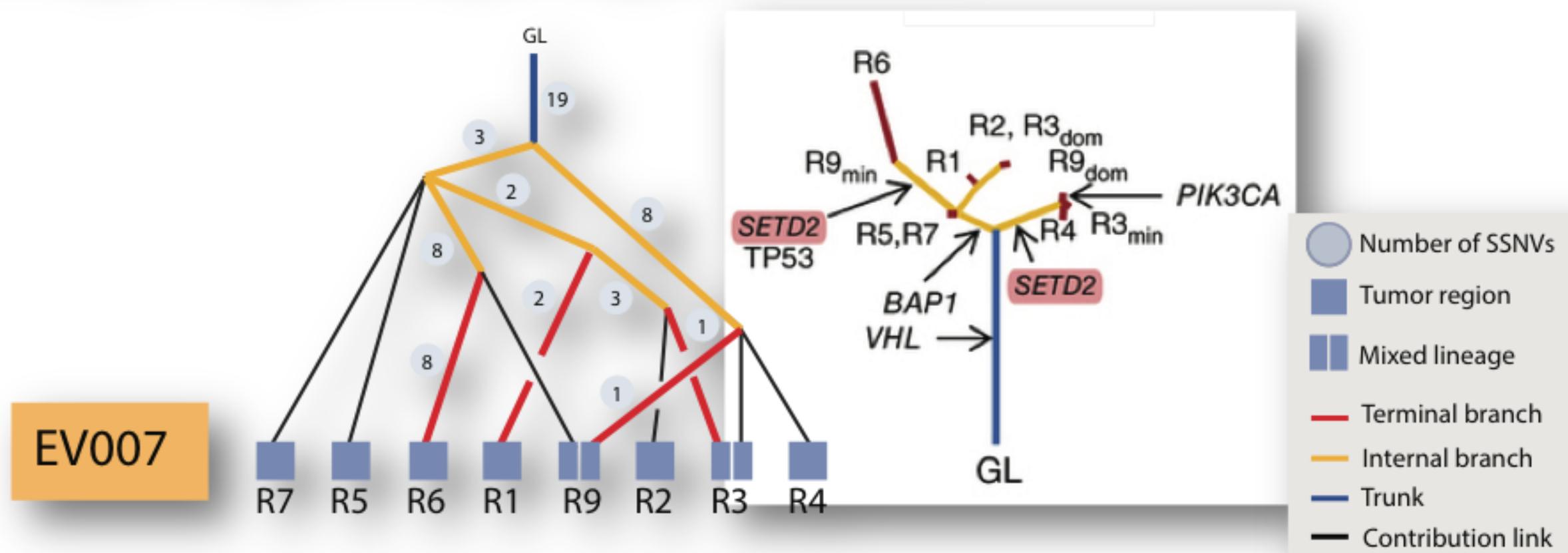
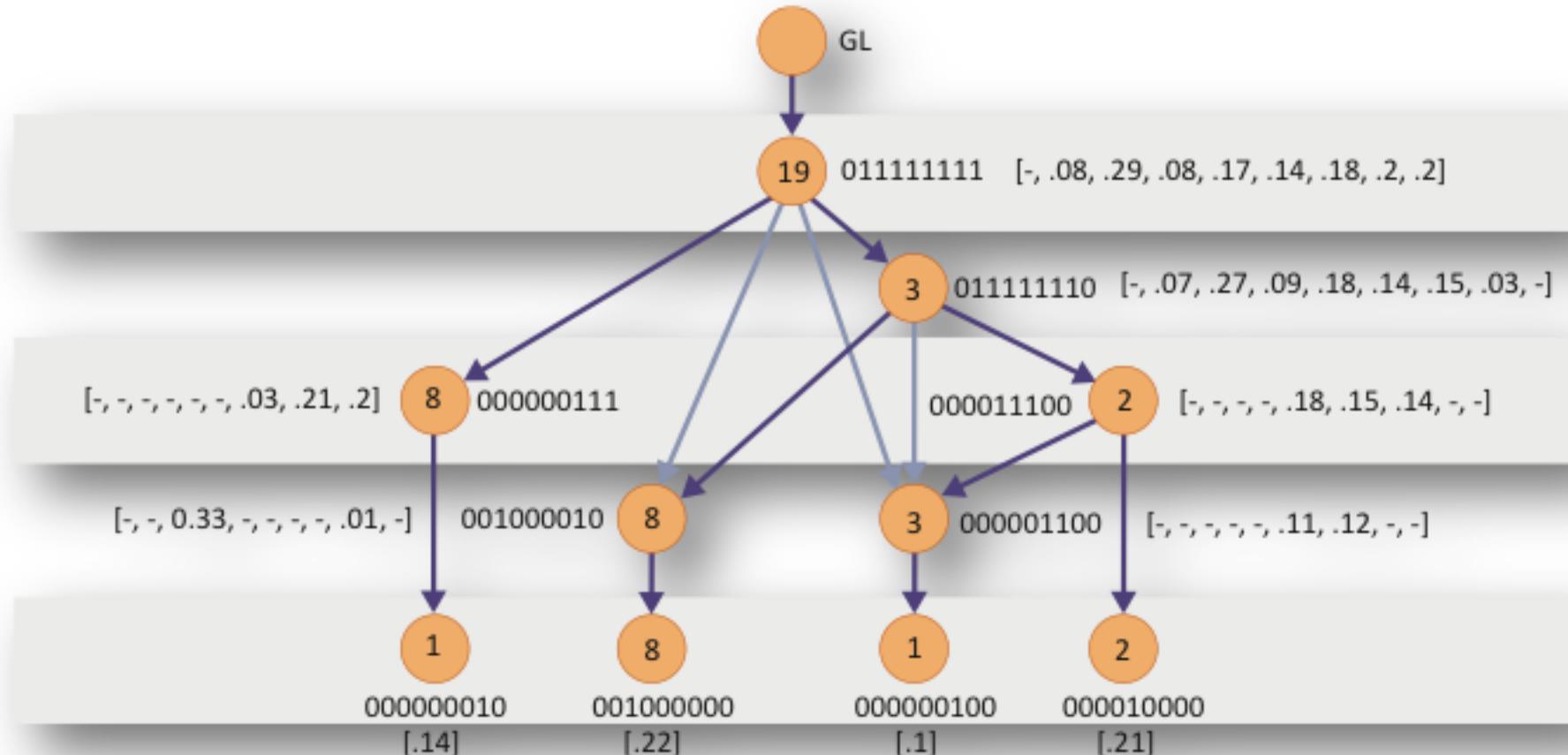


Gerlinger, M., et al. (2014). "Genomic architecture and evolution of clear cell renal cell carcinomas defined by multiregion sequencing." *Nature genetics* **46**(3): 225-233.

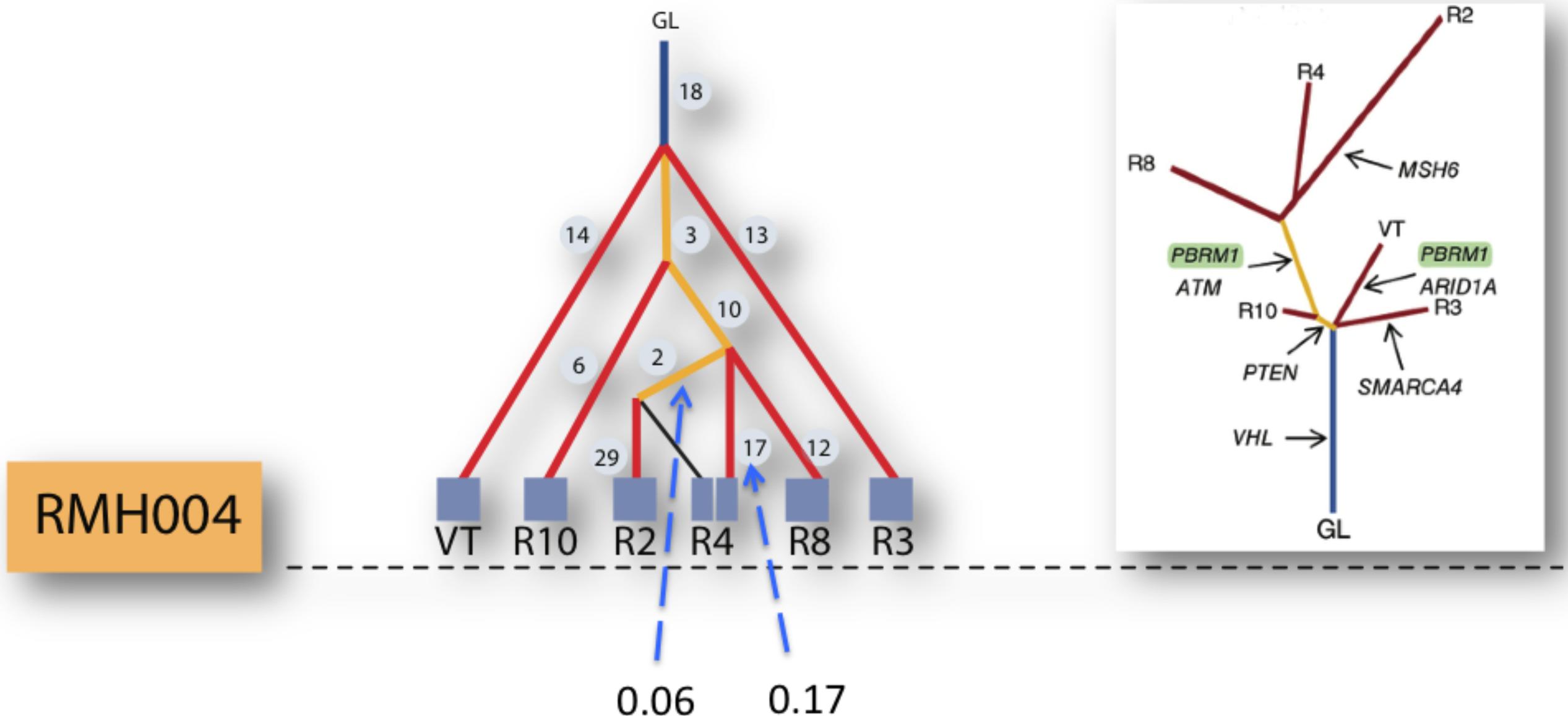
ccRCC Study by Gerlinger et. al (2014)



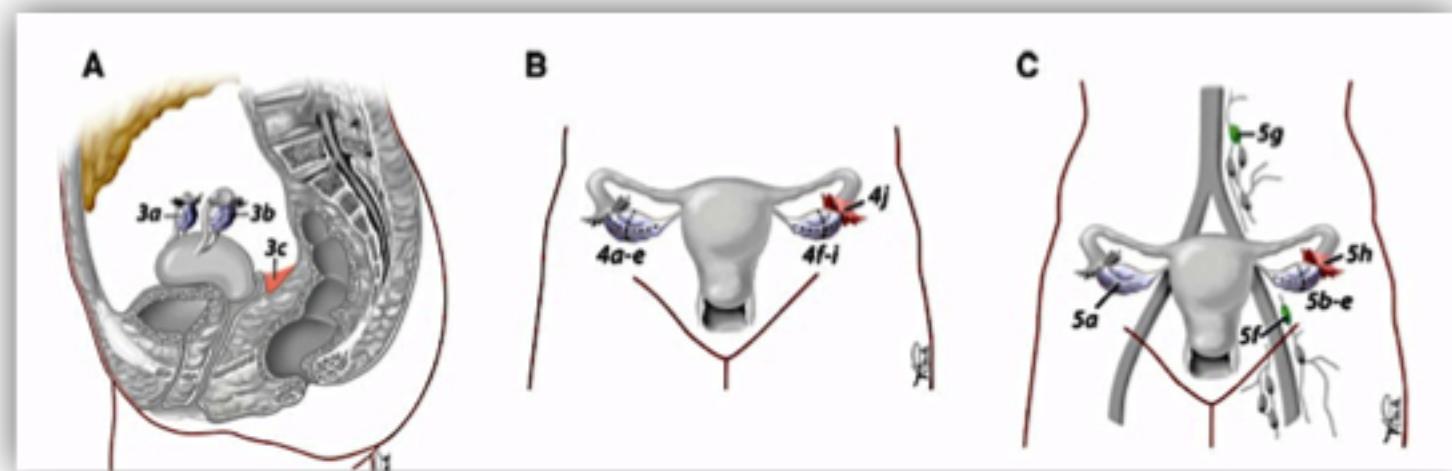
ccRCC Study by Gerlinger *et. al* (2014)



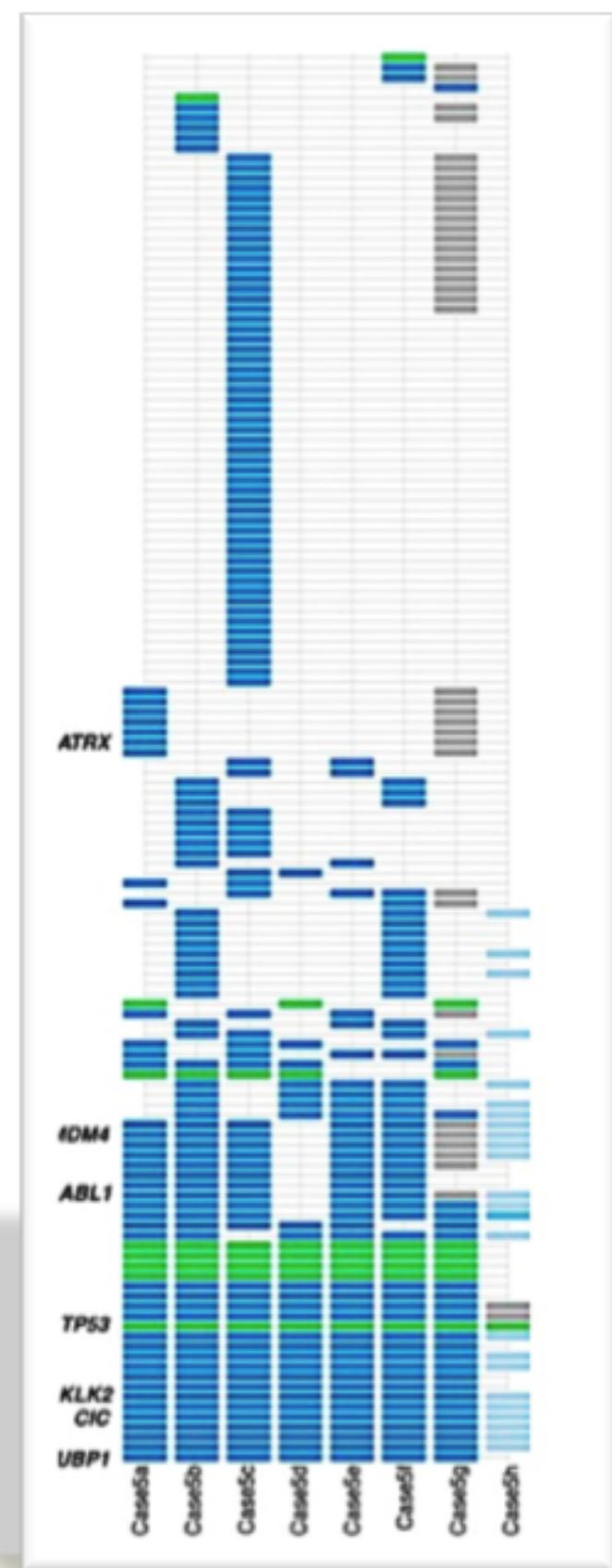
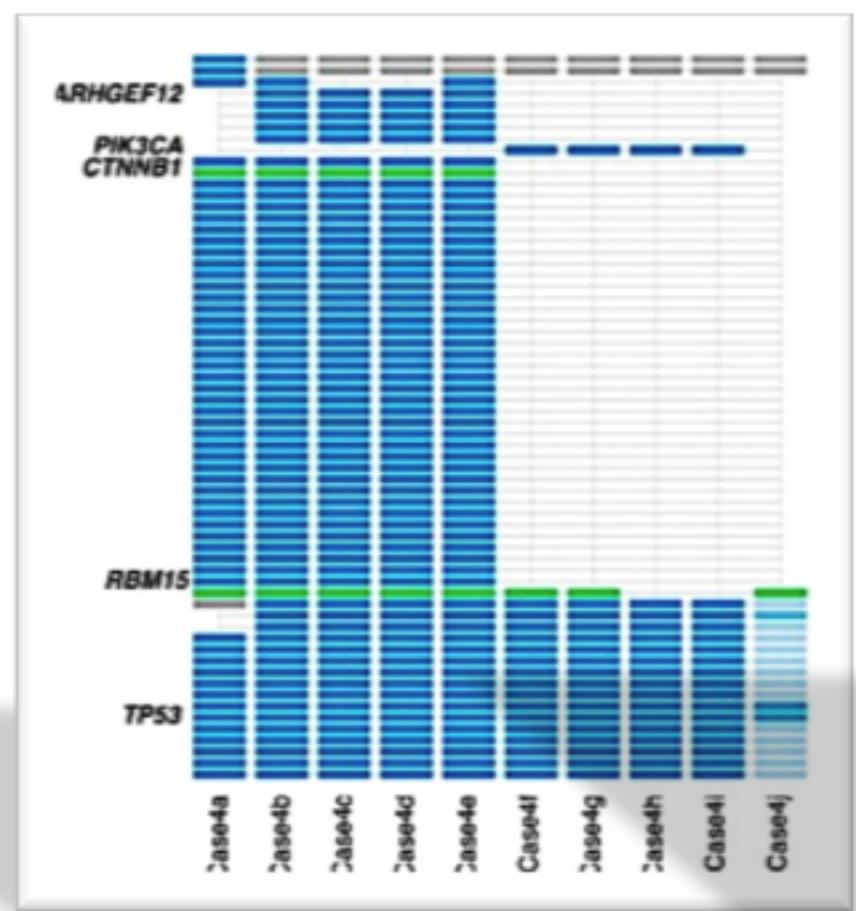
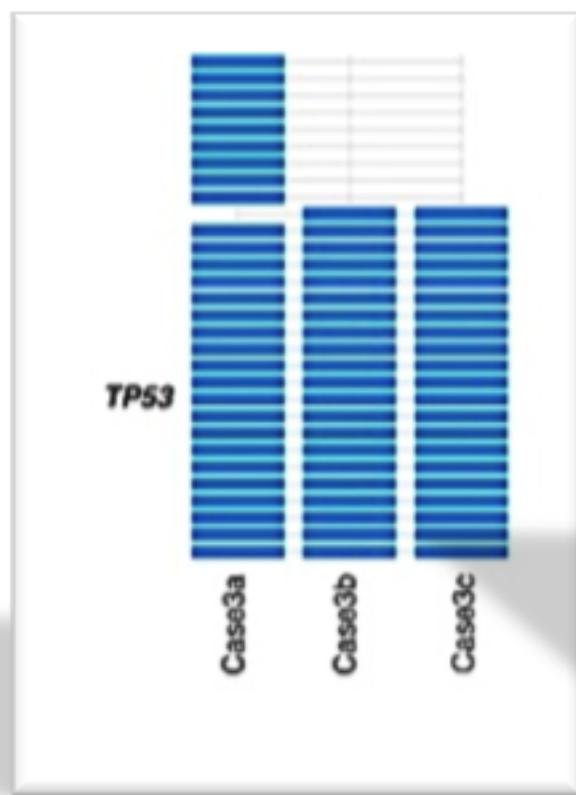
ccRCC Study by Gerlinger *et. al* (2014)



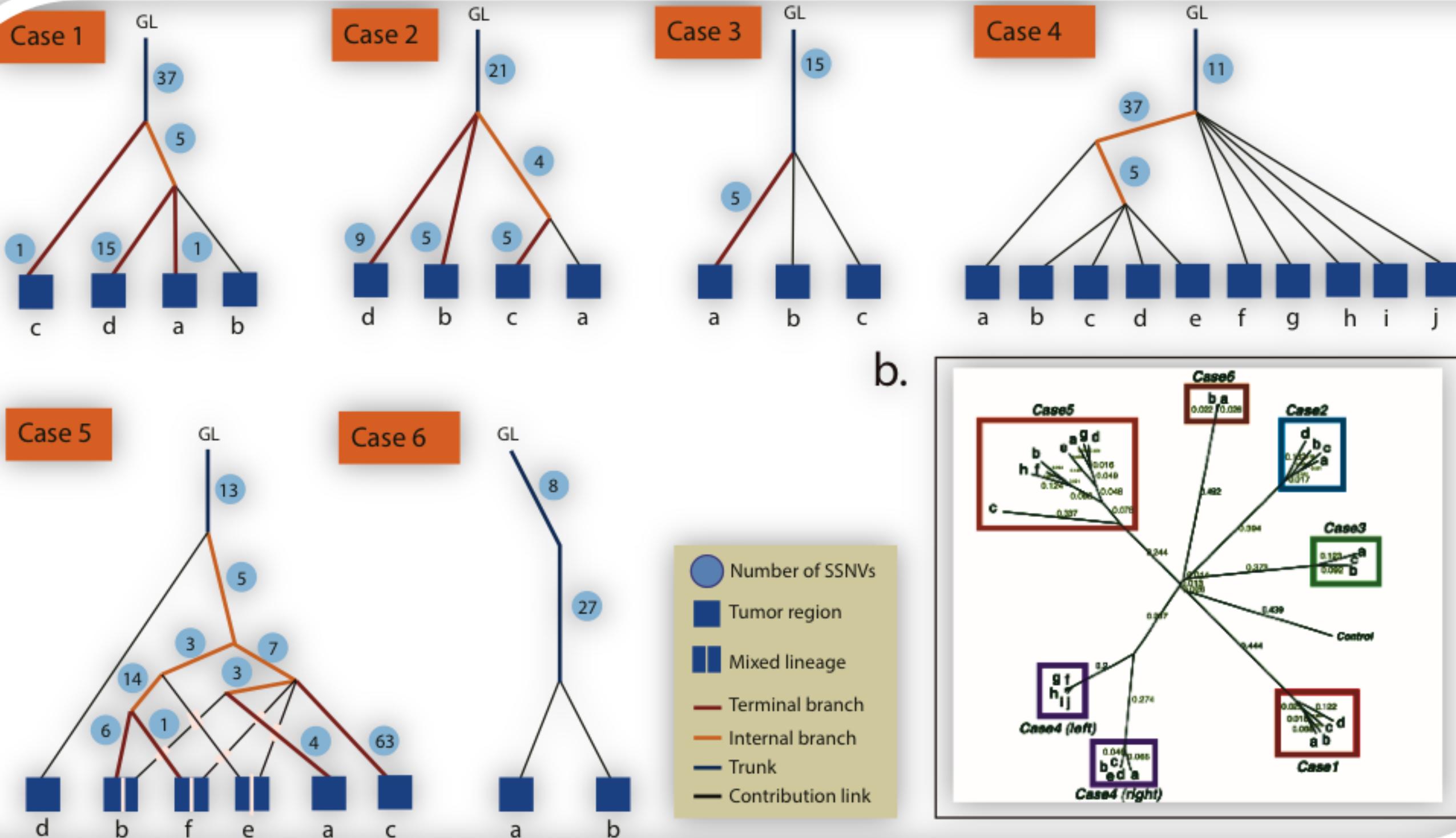
HGSC Study by Bashashati et. al (2013)



19 tumors, 6 patients, 340 SNVs



HGSC Study by Bashashati et. al (2013)



LICHeE Runtime DEMO Movie

```
viq@eflop1:~/srk
```

EDGES:

```
0->4  
0->3  
0->10  
4->11  
4->12  
4->5  
4->7  
4->2  
4->9  
4->2  
3->1  
11->7  
11->8  
12->6
```

Nodes:

	011111111111	011110000000	000000001111	000000111000	001110000000	000000111000	000000001000	000000100000	000000000100	010000000000	000000100000	000000001000	
0	6	0	0.2	0.24	0.22	0.10	0.22	0.10	0.13	0.16	0.11	0.05	0.17
11	21	0	0.19	0.22	0.2	0.10	0	0	0	0	0	0	0
12	3	0	0	0	0	0	0	0	0.15	0.12	0.05	0	0.17
4	4	0	0	0	0	0	0.19	0.14	0.03	0	0	0	0
5	2	0	0	0.01	0.16	0.07	0	0	0	0	0	0	0
7	3	0	0	0	0	0	0.2	0.13	0	0	0	0	0
2	2	0	0	0	0	0	0	0.09	0	0	0	0	0
9	1	0	0	0	0	0	0.07	0	0	0	0	0	0
2	6	0	0	0	0	0	0	0	0	0	0	0.13	0
10	4	0	0.19	0	0	0.2	0	0	0	0	0	0	0
10	10	0	0	0	0	0	0	0.12	0	0	0	0	0
10	1	0	0	0	0	0	0	0	0	0	0	0	0

Found 1 valid trees

best tree error score: 0.06257746445281244

Samples:

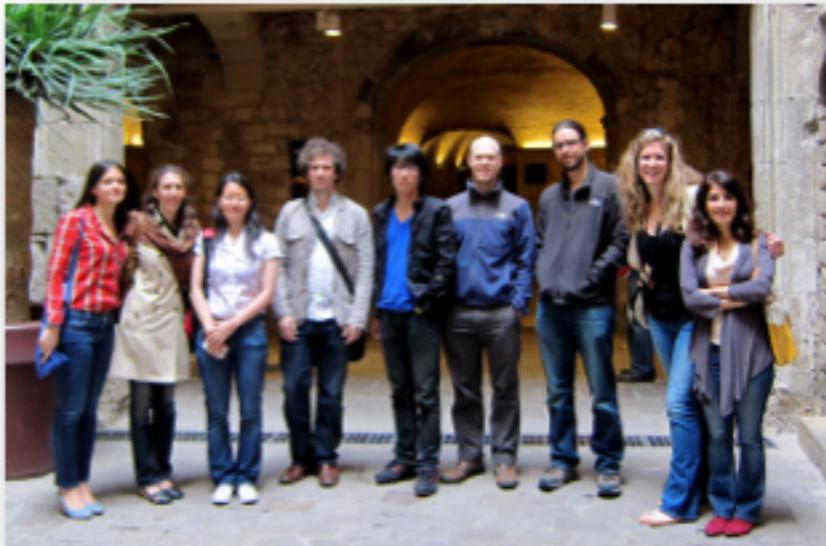
- 0: Normal
- 1: R2
- 2: R3
- 3: R4
- 4: R11
- 5: R10
- 7: R9
- 8: R5
- 9: R6
- 10: R7
- 11: R8

0:b6a23066e:release viq@





Serafim Batzoglou Lab @ Stanford University



Acknowledgements

Raheleh Salari

Iman Hajirasouliha

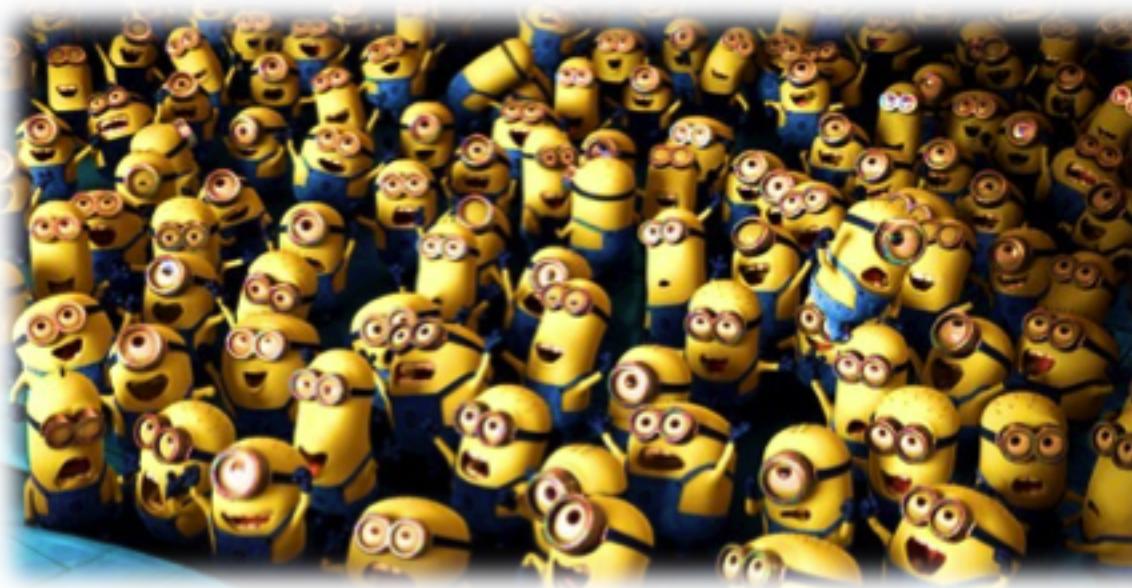
Dorna Kashef-Haghghi

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Robert West

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Serafim Batzoglou



Everyone in the audience

Thank You