

Fighting the Tuberculosis Pandemic using Machine Learning

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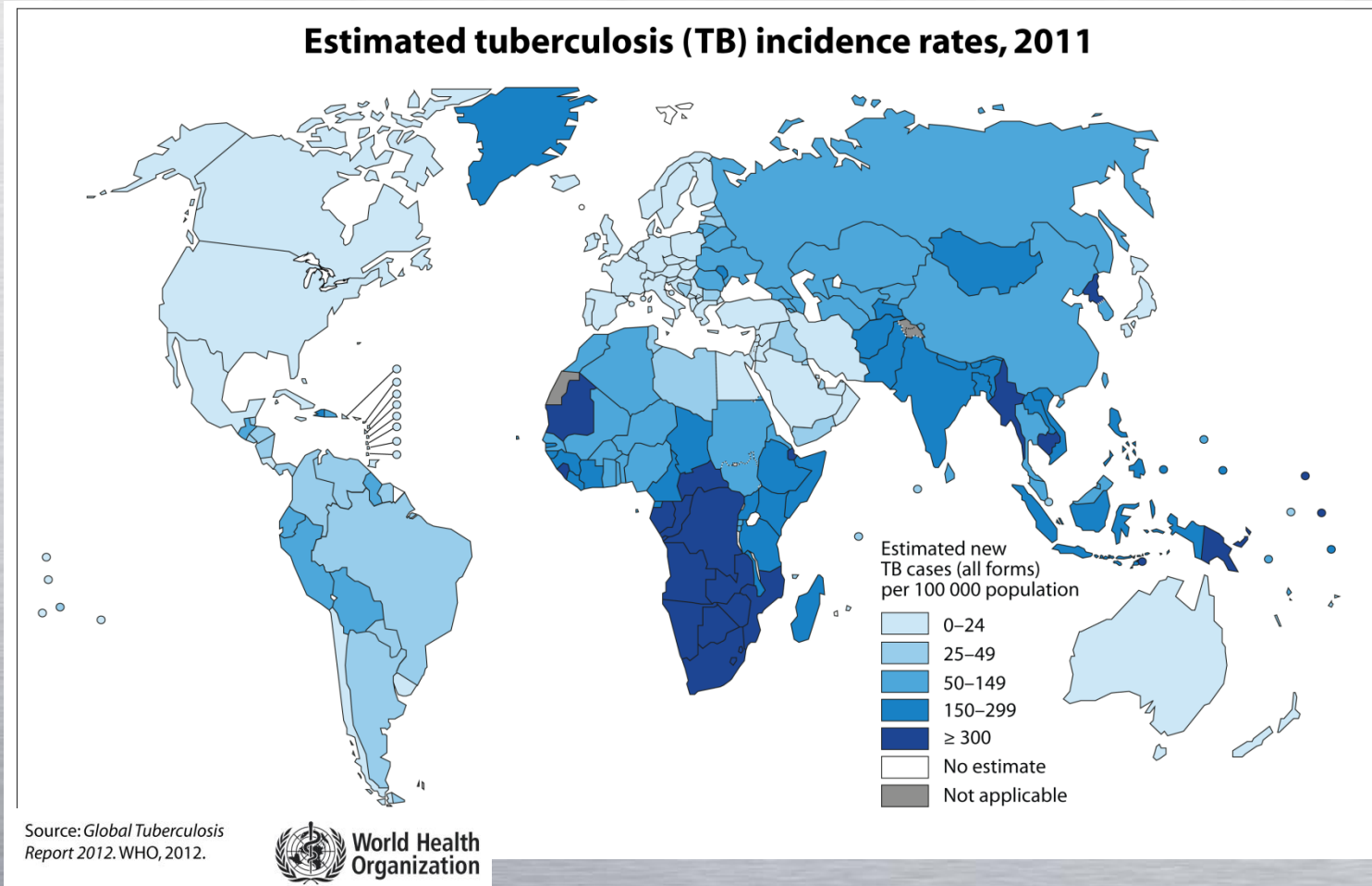
Tuberculosis Tracking and Control

<http://www.cs.rpi.edu/~bennek/tbinsight/index.html>

This work is supported by NIH R01LM009731.

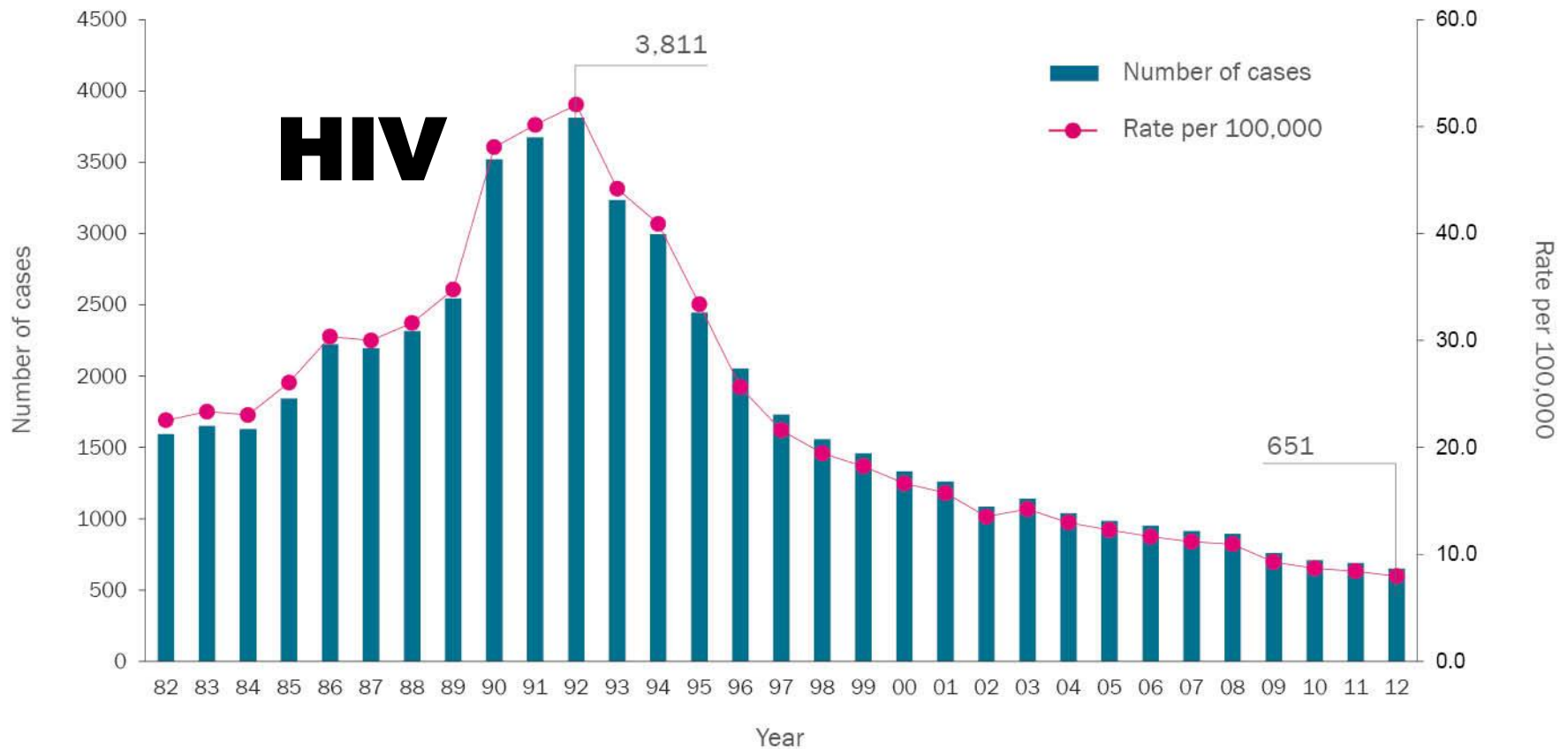


1/3 of World Latently Infected with TB



2.4 Million Deaths per Year

Tuberculosis cases and rates, New York City, 1982-2012



1. Rates are based on official Census data.

Number of TB Cases in U.S.-born vs. Foreign-born Persons United States, 1993–2011*



*Updated as of June 25, 2012.



Drug Resistance Threat

Countries that had notified at least one case of XDR-TB by the end of 2011



Division of
of any
boundaries.
to be full agreement.

Source: *Global Tuberculosis Report 2012*. WHO, 2012.



- Susceptible
- Drug Resistant
- MDR-TB - Multi-Drug Resistant
- XDR-TB - Extremely-Drug Resistant
- TDR-TB?- Totally Drug Resistant



Modern TB Control

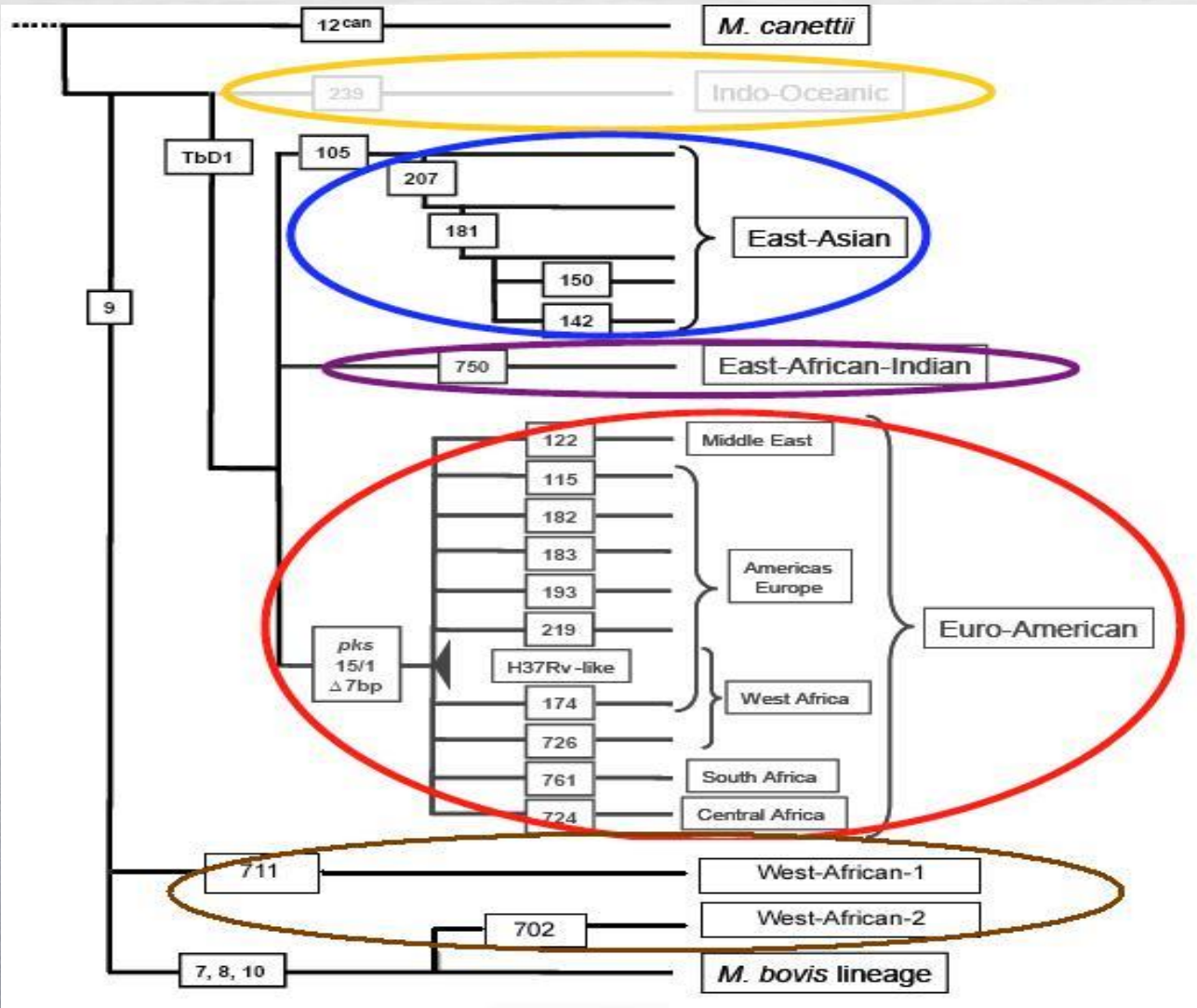


TB Controller: Find source(s) of infection in order to identify people who need treatment and stop future transmission.

Tools: **Contact Investigation**

DNA Fingerprints of TB bacteria

Major Phylogeographic Lineages of the MTBC

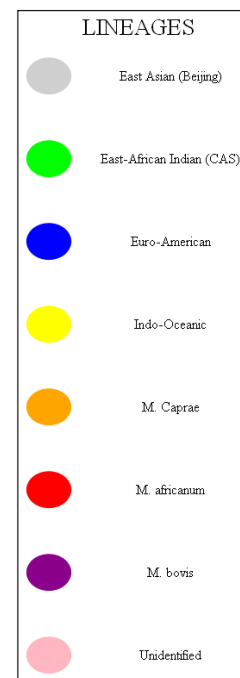
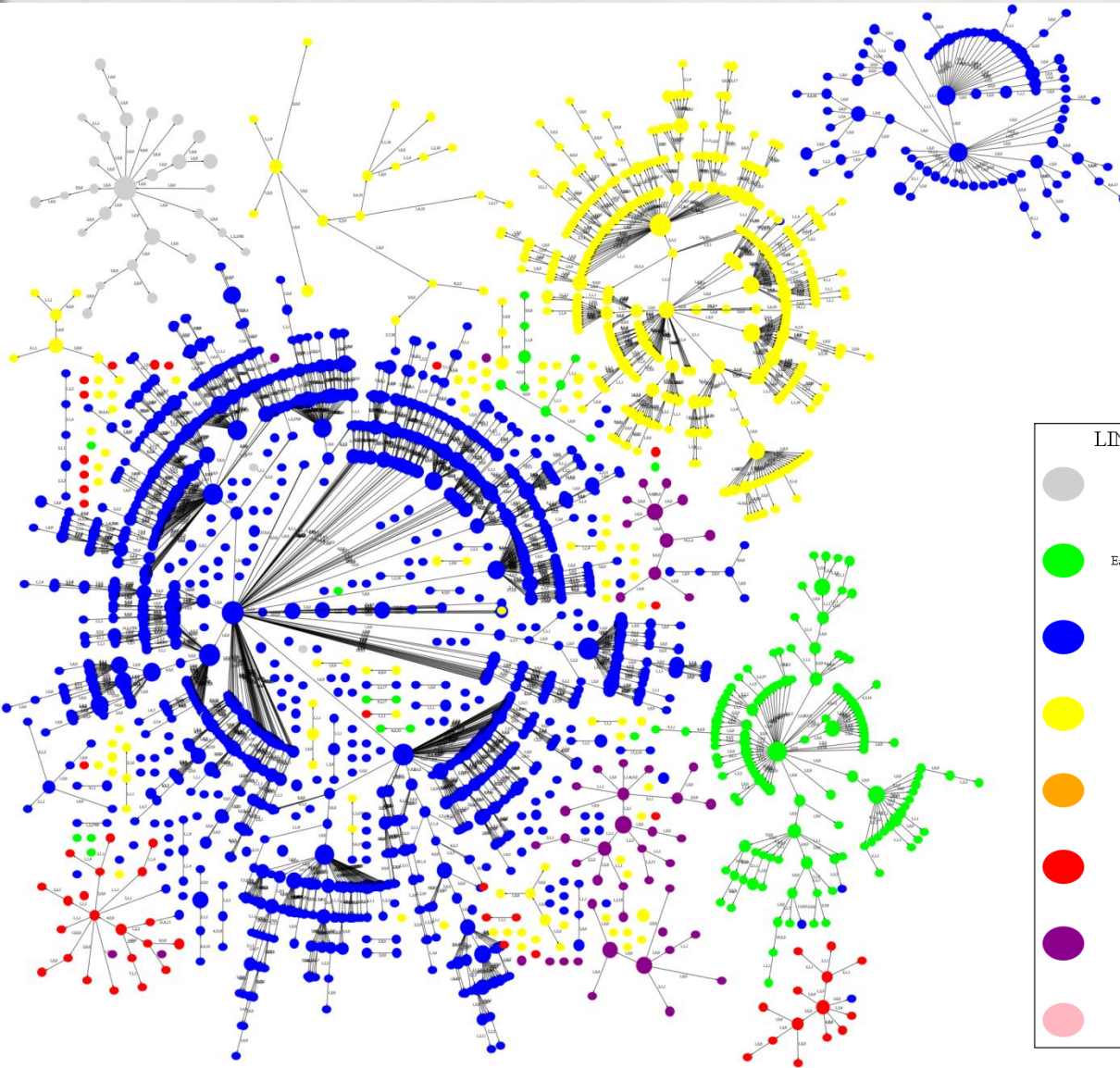


Determined by RD's

Predictable by
Spoligotype

Spoligotype Genetic Diversity 37K Patients in US – 2004-2008

Spoligoforests
labeled by
CDC Expert
Rules



Classification Models

TB-Lineage

- Rule-based model: 2012.
- Bayesian Network: 2010.

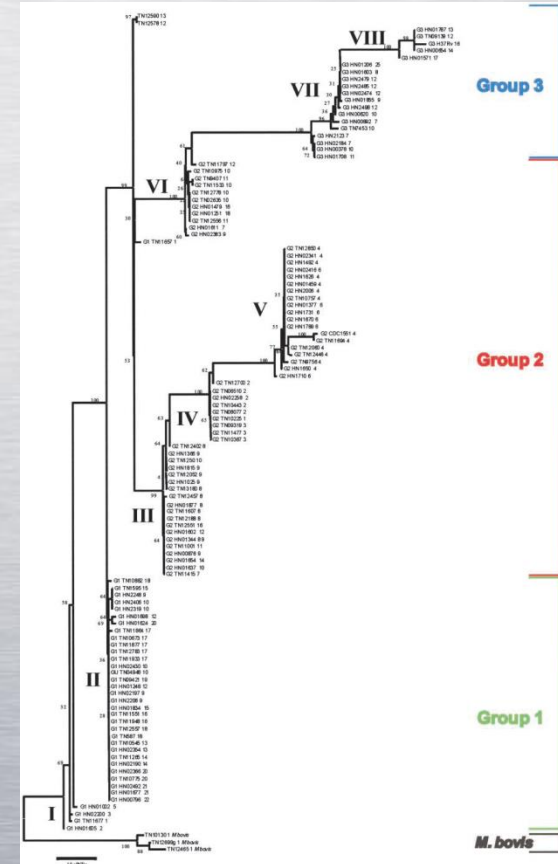
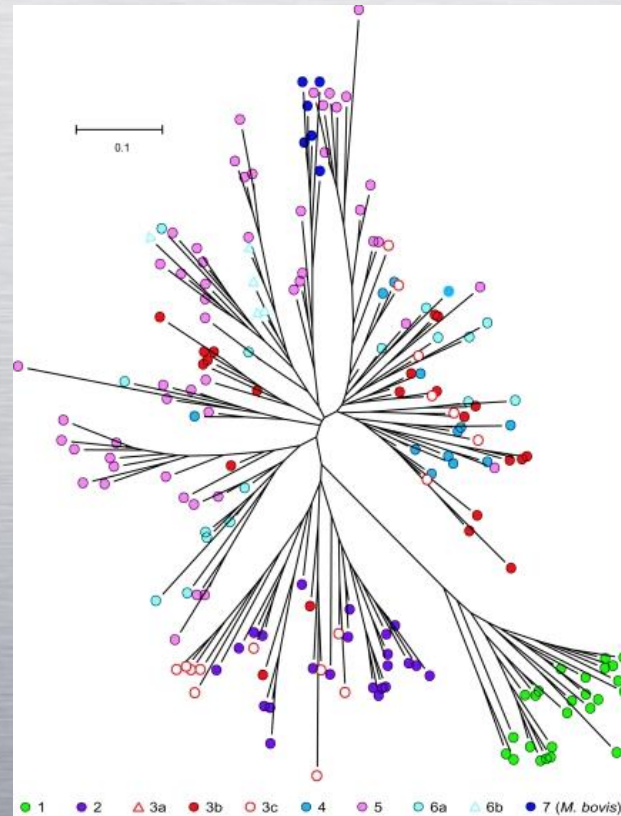
Sublineages-varying opinions

62 sublineages based on spoligotype signatures
[Brudey et al BMC Microbiol. 2006]

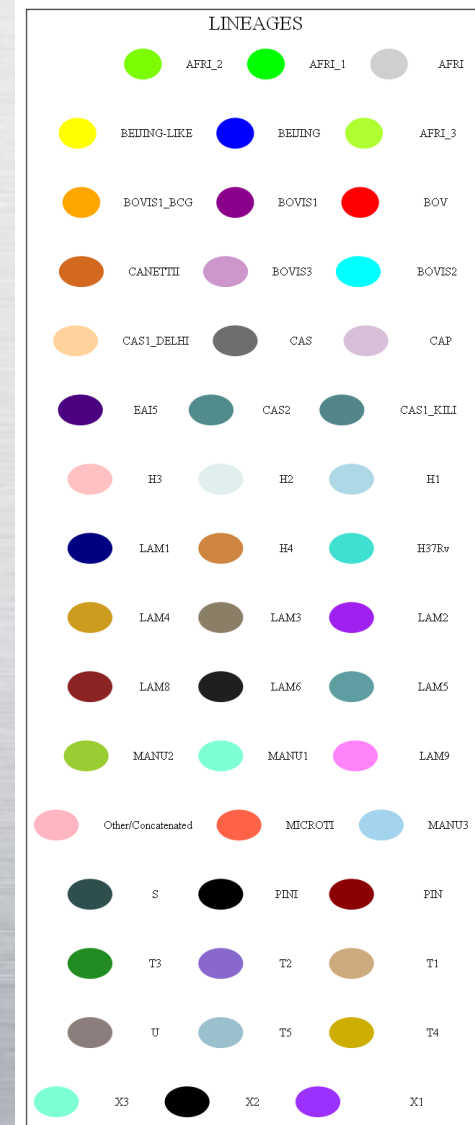
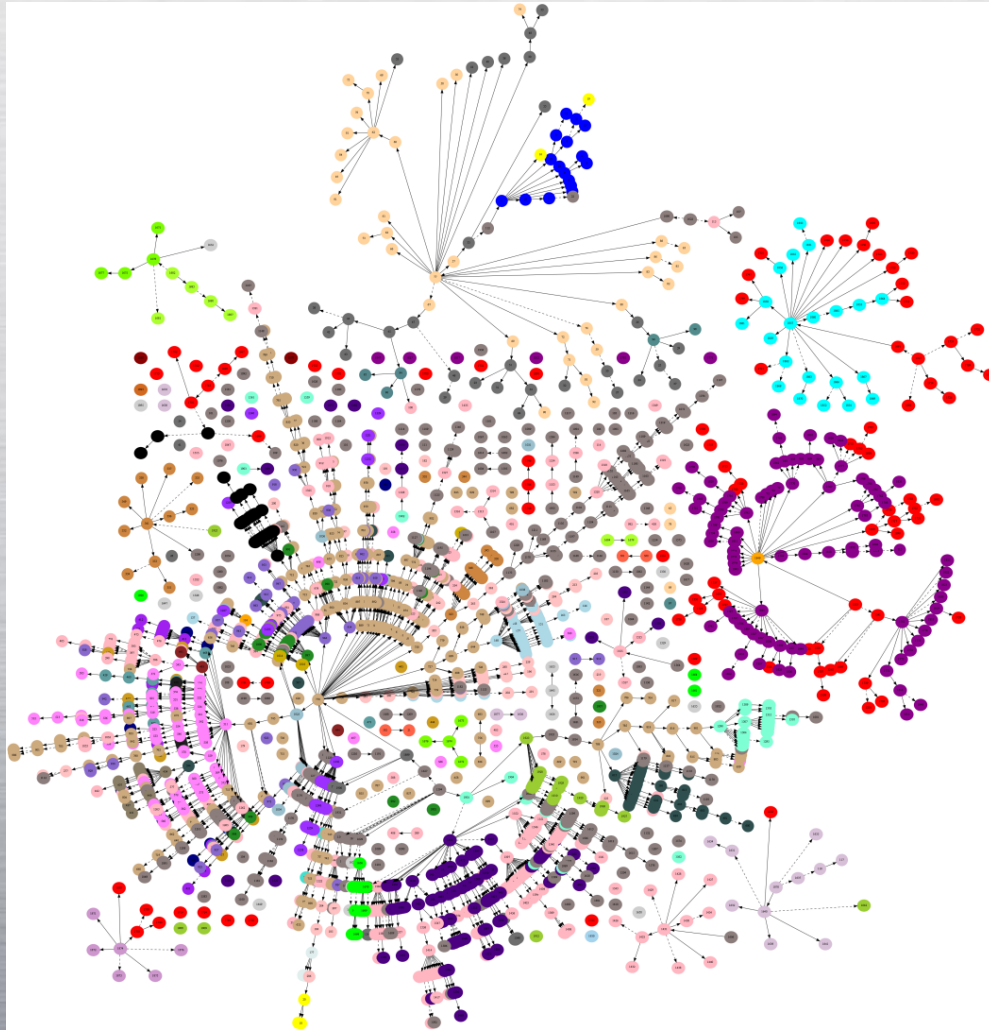
10 groups based on 212 single nucleotide polymorphism (SNP) markers
[Filliol J. Bacteriol 2006]

8 groups based on 230 sSNPs [Gutacker Genetics 2002]

ST n°	lineage/sublineage	Binary spoligotype	Octal**	Total	Frequency
181	AFR1	7107777777771	270	0.75
321	AFR2	7407767777771	74	0.20
430	AFR3	7740777777771	30	0.08
1	BeBrn	0000000000000	4081	11.28
482	BOV1	6767777777700	2401	6.68
683	BOV2	6640777777700	2770	7.71
479	BOV3	6400777777700	204	0.57
647	BOV4	3000007777700	166	0.46
491	BOV1-variant1	6765777777700	170*	0.48*
645	BOV1-variant2	6767777777700	120*	0.33*
676	BOV2-variant1	664077774037000	177*	0.49*
478	BOV2-variant2	66407777763000	426*	1.19*
392	CanE	0000000011100	8	0.03
28	CAS1-GenE	7037777000771	1076	3
21	CAS1-GenE	703776000771	158	0.44
28	CAS1-variant1	7037774000771	128*	0.35*
25	CAS2	7037777000771	79	0.22
495	Zen	7177770000111	49	0.05
100	MANU1	7177777777771	75	0.21
54	MANU2	7177777777771	160	0.45
173	MANU3	7177777777771	188	0.58
539	meso	0000000000000	38	0.11
803	PM1	0740000777700	11	0.03
537	PM2	0000000777700	30	0.08
236	EAI-5	7177777741371	1014	2.82
48	EAI-5CM	7177777741371	450	1.25
19	EAI2-Mantle	6777774741371	544	1.51
39	EAI2-Northeast	6766000041371	72	0.2
11	EAI-IND	4177777741371	432	1.2
199	EAI4-VNM	7177777441371	347	0.97
591	EAI6-RO1	717777741371	169	0.47
1008	EAI7-RO2	7177777000371	68	0.19
169	EAI8-RO3	417777741371	83	0.17
47	H1	7177777402071	1206	3.44
62	H1-variant1	7177777402071	171*	0.46*
2	H2	6000000000071	298	0.82
59	H3	7177777777771	1963	5.62
127	Jca	5177777420771	374	1.04
28	LAM1-GenE	6777760760771	484	1.35
17	LAM2	7177750760771	411	1.14
33	LAM3	7817760760771	509	1.42
41	LAM4	7177760760771	146	0.41
93	LAM5	7177760760771	264	0.73
64	LAM6	7177760760771	172	0.48
141	LAM7-GenE	7177760760771	160	0.45
290	LAM8	7177760760771	22	0.06
42	LAM9	7177760760771	1930	5.37
61	LAM10-GenE	717777760760771	432	1.2
59	LAM11-GenE	717776060760771	240	0.67
209	LAM12-GenE	7807760760771	24	0.07
11	T1	7177777776771	6412	17.65
280	T1-RUS2	7700077776771	63	0.18
34	T2	7670777776771	648	1.8
82	T2	7177777776771	840	2.34
139	T2-Uganda	7177777776771	48	0.13
15	T3	7177777776771	277	0.77
146	T3-ETH	7177000776771	221	0.62
46	T4	7177777776771	168	0.47
14	T4-CEU1	7177777776771	133	0.37
44	T5	7177777776771	298	0.83
98	T5-Andres	7177777776771	103	0.29
24	T5-RO1	7177760776771	100	0.28
173	T5-GenE	7177000776771	11	0.03
41	H31-GenE	7177777776771	147	0.41
627	T3-OSA	7477377776771	19	0.05
119	X1	7177777776771	956	2.66
137	X2	7177777776001	1268	3.62
38	X2-variant1	7177640760001	125*	0.34*
92	X3	7000777776771	899	1.96
103	X3-variant1	7000007776771	141*	0.39*
197	X3-variant2	7000977760771	244*	0.67*
4	unsubstantiated	7177777776700	158*	0.43*
1	LAM9S	00000000760771	119*	0.32*
73	T3-variant	71777777760771	154*	0.29*
210	unsubstantiated	77867777760771	104*	0.29*
40	X3-variant	70000077760771	102*	0.29*
46	unsubstantiated	7177777700000	95*	0.26*



What's the story with sublineages?



Spoligoforest – drawn using GraphViz Twopi
Data labels –SpolDB4 [Brudey et al 2006]

SPOTCLUST: Hidden-Parent Bayesian Network for Spoligotypes (Vitol et al, 2006)

Standard resource for **tuberculosis sublineage identification** used in over 96 publications.

Strains of *Mycobacterium tuberculosis* from Western Maharashtra, India, Exhibit a High Degree of Diversity and Strain-Specific Associations with Drug Resistance, Cavitory Disease, and Treatment Failure

Characterization of multiple and extensively drug resistant *Mycobacterium tuberculosis* isolates with different ofloxacin-resistance levels

Distinct clinical and epidemiological features of tuberculosis in New York City caused by the RD^{Rio} *Mycobacterium tuberculosis* sublineage

■ *Mycobacterium bovis* infection in livestock workers in Ibadan, Nigeria: evidence of occupational exposure

High prevalence of subclinical tuberculosis in HIV-1-infected persons without advanced immunodeficiency: implications for TB screening

Whole cell & culture filtrate proteins from prevalent genotypes of *Mycobacterium tuberculosis* provoke better antibody & T cell response than laboratory strain H₃₇Rv

Rule-Based Method: (Sub)Lineage Visual Rules

M. africanum 

Haarlem 2 

- Determined by human experts.
- Ill-defined.
- Incomplete.
- Frequently ambiguous.
- No precedence.
- May or may not correspond to actual evolutionary groups.

First Try – Naïve Bayes

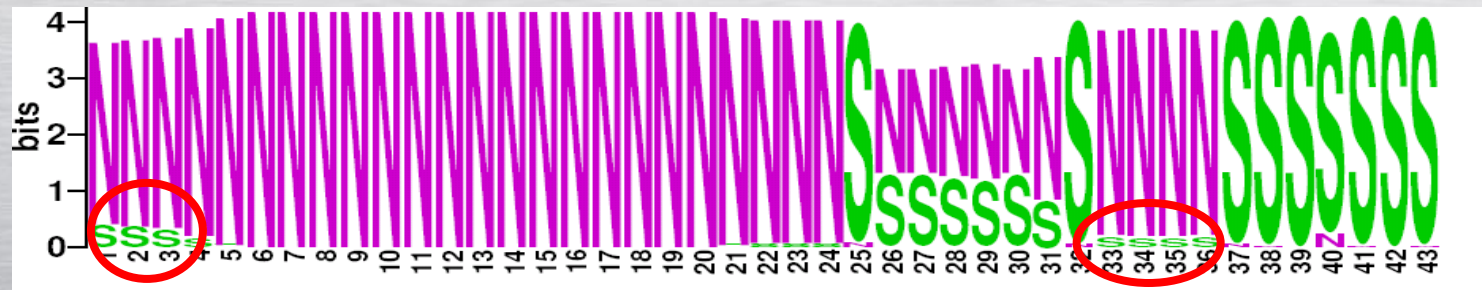
M. tuberculosis Haarlem2 Family

- Prototype = probabilities



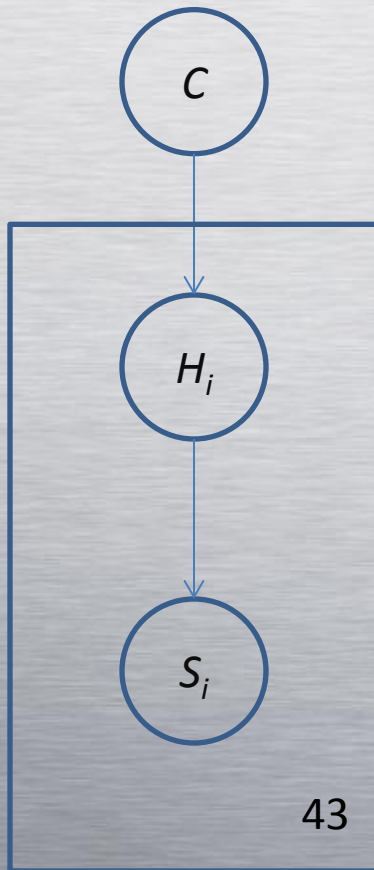
- Bernoulli Mixture Model

S - spacer present
N - spacer absent



- Biology is wrong!!!

Unsupervised Hidden Parent Multivariate Bernoulli Mixture Model

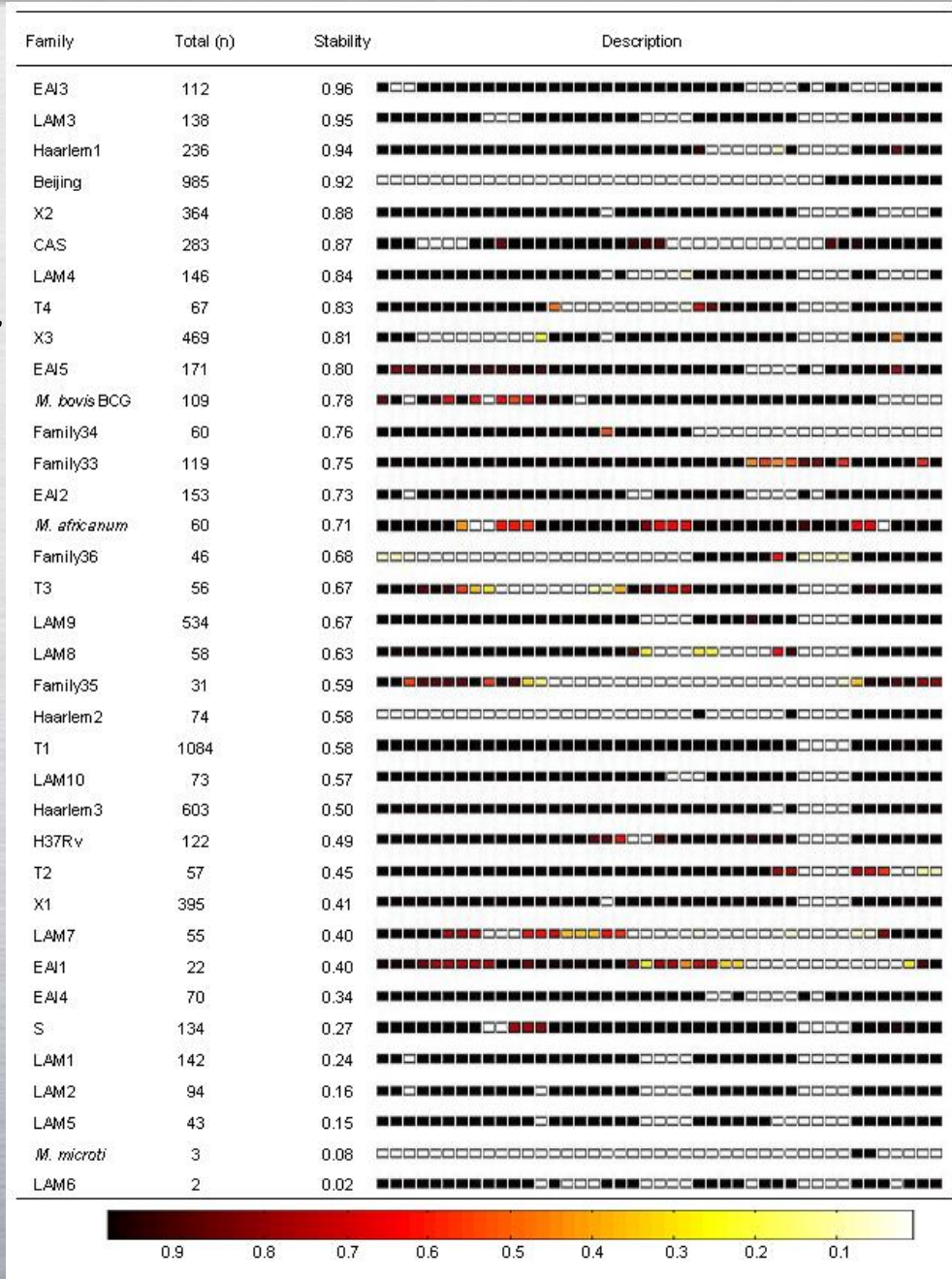


Model child spacer S , given unobserved parent spacer H

- With very high probability child matches parent
- Children are much more likely to lose spacer than gain
 - $P(S=1|H=1) = 0.99$,
 $P(S=0|H=1) = 0.01$
 - $P(S=0|H=0) = 1-1e-7$
 $P(S=1|H=0) = 1e-7$

SPOTCLUST (2006)

- Unsupervised except
34 SPOLDB2 Visual Rules
used to initialize clusters
- Trained using 535
spoligotypes
- Number of
sublineages (36)
picked by MCCV



New Challenges

- More data: 119,684 isolates from US CDC, NYDOH, NY State DOH, and Institut Pasteur de Guadeloupe, MIRUVNTRPlus
- More types of DNA fingerprints – Spoligotypes and MIRU
- More proposed sublineages (70?)
- Putative labels from multiple experts
- Missing Data

Who's Right?



Ctop	Cmid	Csub
Indo-Oceanic	Bangladesh	EAI6-BGD1 EAI7-BGD2
	India	EAI3-IND
	Manila	EAI2-Manila
	Mexico	EAI-Mexico
	Nonthaburi	EAI2-nonthaburi
	Vietnam	EAI4-VNM
	Unknown Mid-level	EAI1-SOM EAI2 EAI8-MDG
<i>Mycobacterium africanum</i>	West African 1	AFRI_2 AFRI_3
	West African 2	AFRI_1
<i>Mycobacterium bovis</i>	<i>Mycobacterium bovis</i>	BOV_1 BOV_2 BOV_3
<i>Mycobacterium canettii</i>	<i>Mycobacterium canettii</i>	Canettii
<i>Mycobacterium caprae</i>	<i>Mycobacterium caprae</i>	Caprae
<i>Mycobacterium microti</i>	<i>Mycobacterium microti</i>	Microti
<i>Mycobacterium mungi</i>	<i>Mycobacterium mungi</i>	M. mungi
<i>Mycobacterium pinnipedii</i>	<i>Mycobacterium pinnipedii</i>	Pini1
		Pini2

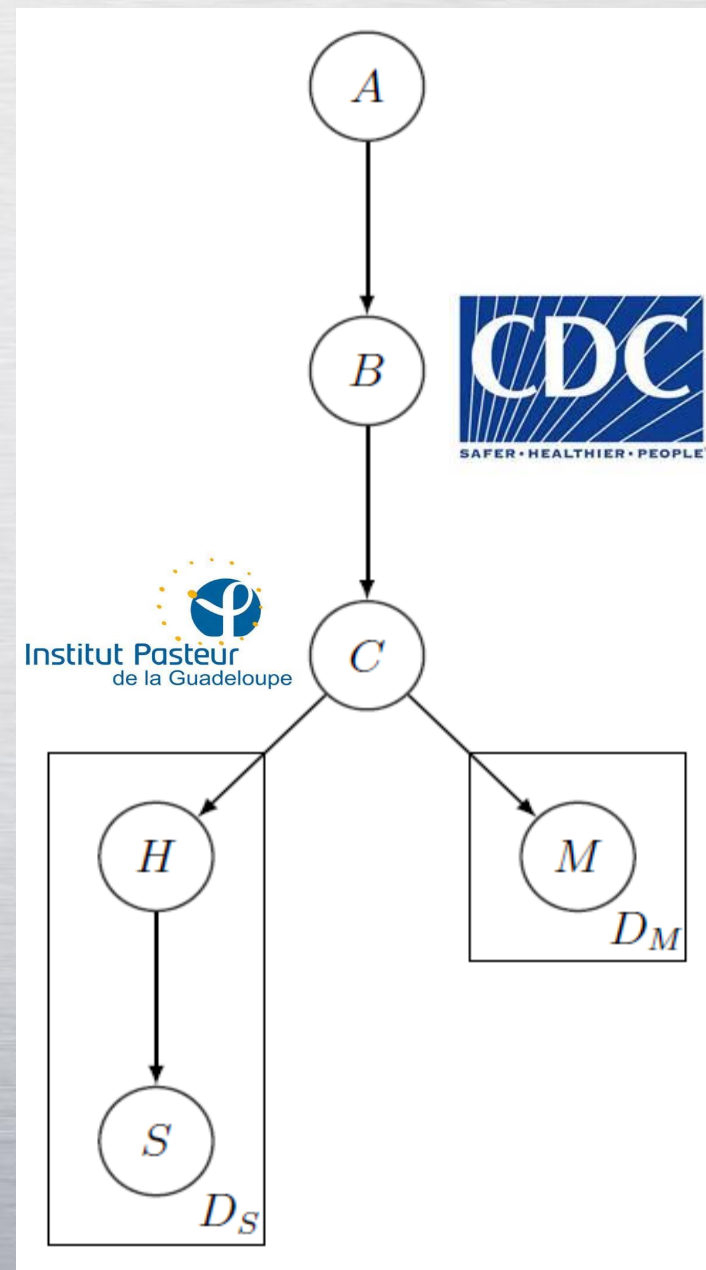
Semi-supervised Hierarchical Lineage Model

A: 12 Major Lineages

B: 22 Mid-level Lineages

C: 70 + 9 Sub-lineages

- Estimated 92% Cross-validated Accuracy



Major Lineage Results

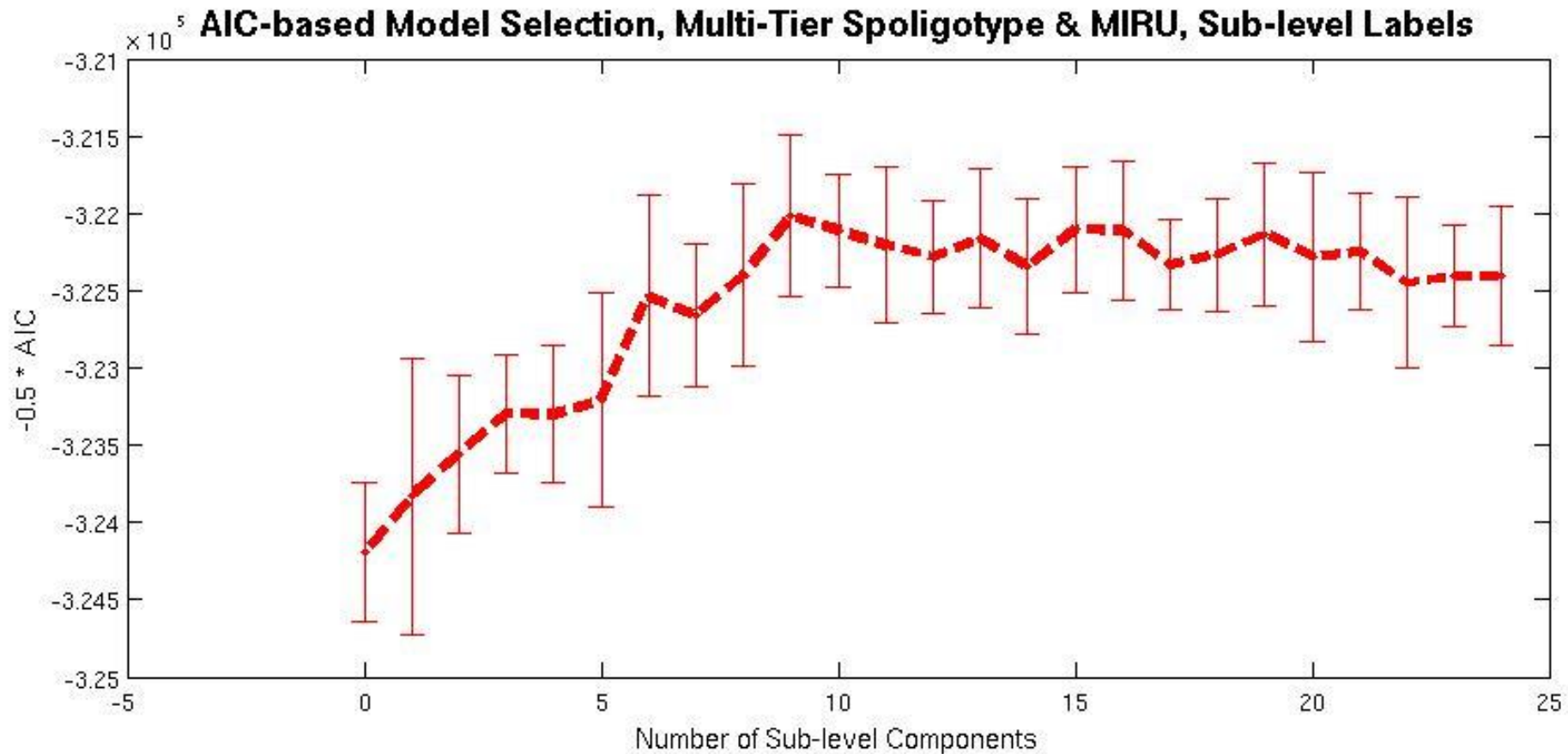
- Balanced Classification Rate about 98%
- No changes in major lineages
- MANU Modern?

Top Pred. Label	Spoligotype Probabilities
M. africanum	
M. bovis	
M. canettii	
M. caprae	
M. microti	
M. mungi	
M. pinnipedii	
East Asian (Beijing)	
East-African Indian	
Euro-American	
Indo-Oceanic	
Manu	

Top Pred. Label	MIRU2	MIRU4	MIRU10	MIRU16	MIRU20	MIRU23	MIRU24	MIRU26	MIRU27	MIRU31	MIRU39	MIRU40
M. africanum												
M. bovis												
M. canettii												
M. caprae												
M. microti												
M. mungi												
M. pinnipedii												
East Asian (Beijing)												
East-African Indian												
Euro-American												
Indo-Oceanic												
Manu												

MIRU Legend: 0 1 2 3 4 5 6 7 8 9 A R S T U V W X Y Z

Model Adds Sublineages but not Mid or Major Lineages

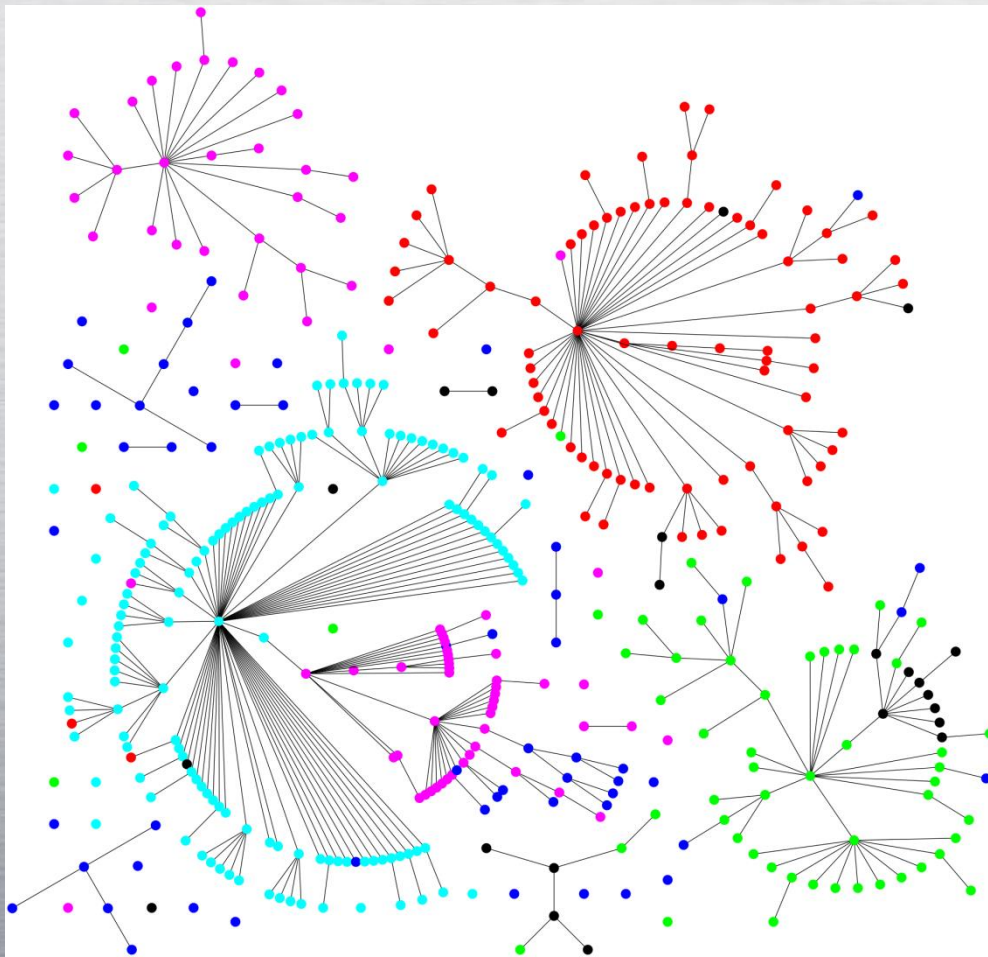


Ten New Putative Sublineages

- Discovers new sublineages and assigns them mid-level lineage
- New lineage characterized by “long” deletions frequently covering “typical” lineage deletions
- Covers many previously unlabeled isolates.

X	0.161	X1	0.362	4145	
X	0.161	X2	0.371	4282	
X	0.161	X3	0.202	2574	
X	0.161	LAM8	0.015	318	
X	0.161	OtherSub5	0.01	164	
X	0.161	OtherSub7	0.033	534	

Visualizing X Lineage Spoligoforest



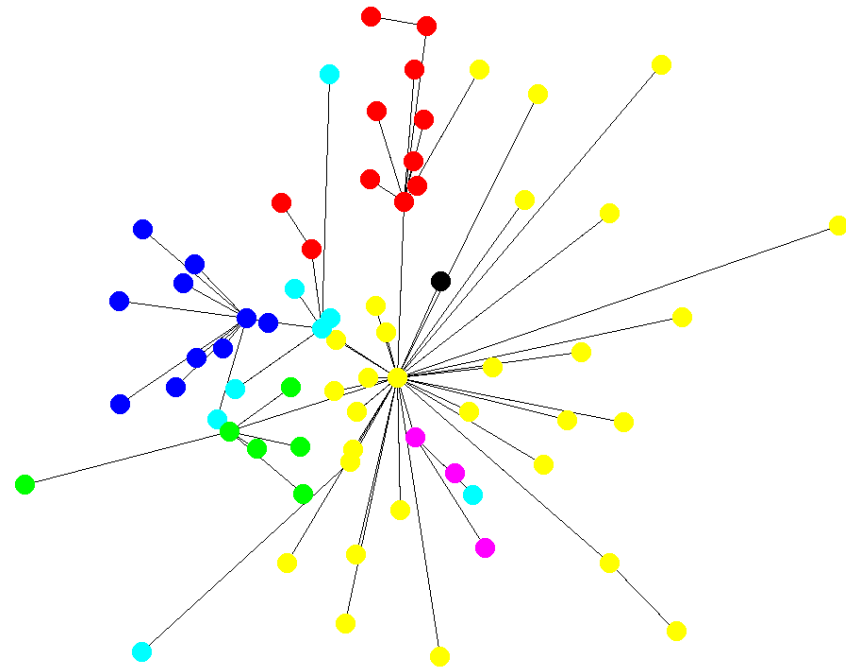
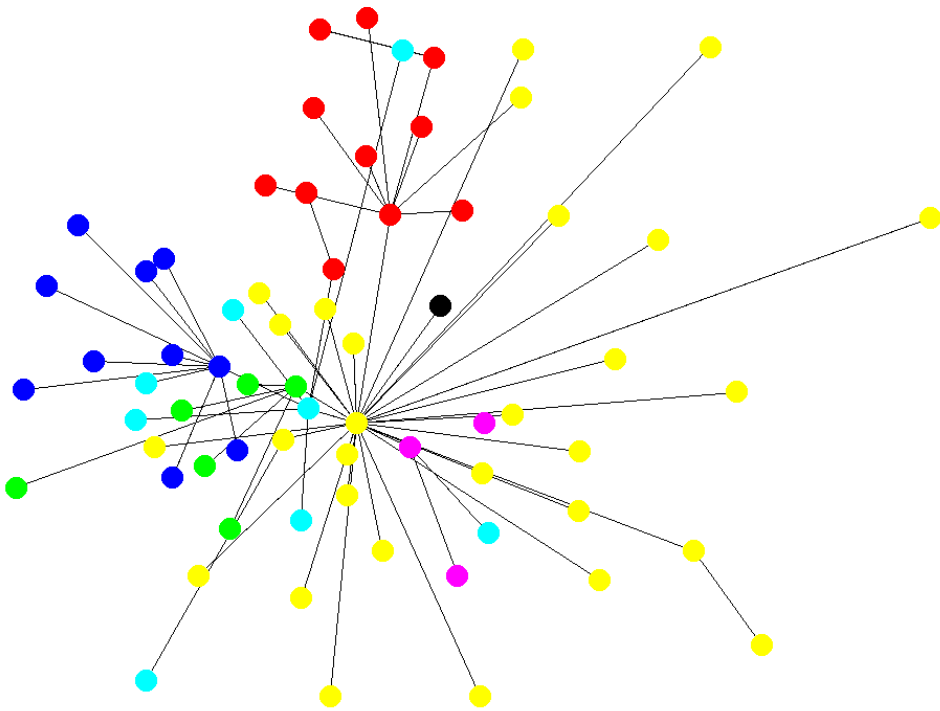
- LAM8
- OtherSub5
- OtherSub7
- X1
- X2
- X3

Graphvis Two-PI

Multi-Objective Embedding Methods

Good visualizations

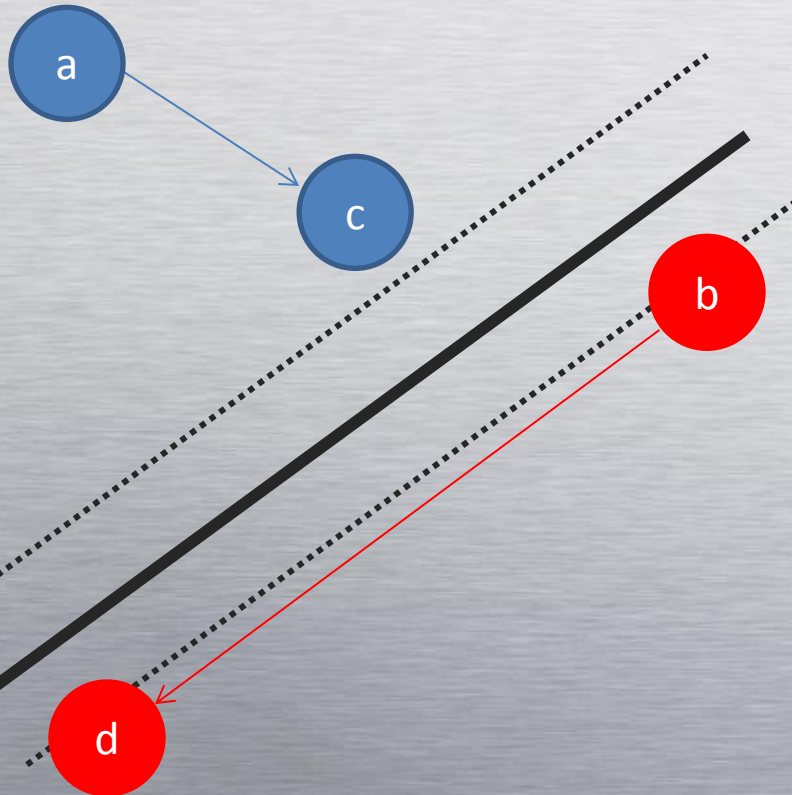
minimize *stress* and *edge crossings*



Edge Crossing Constraints as Classification

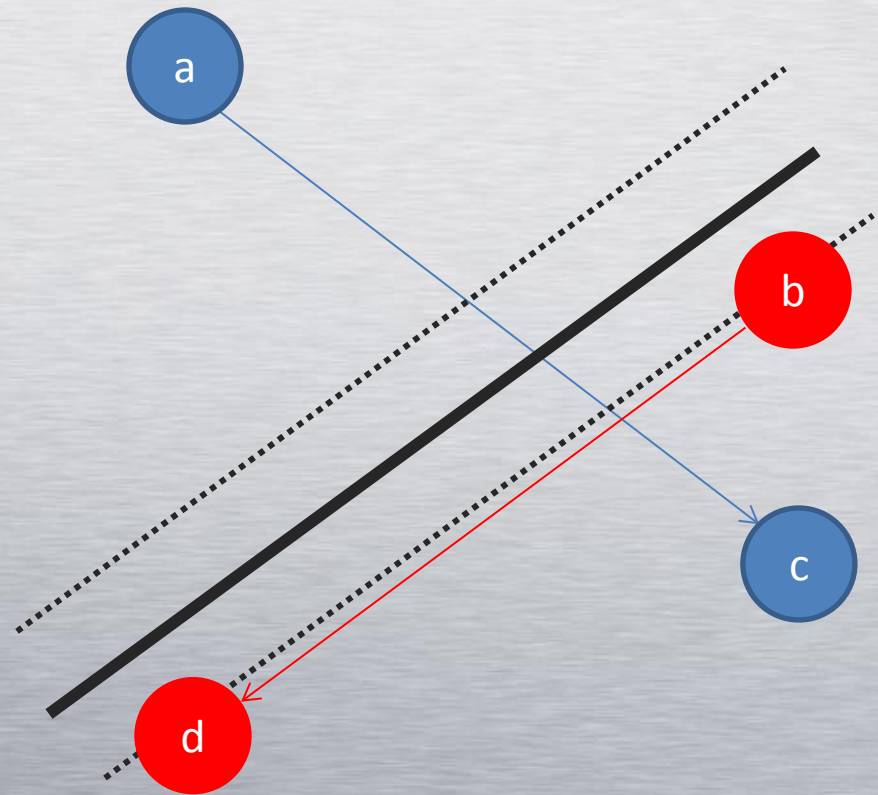
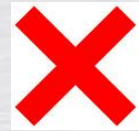
No Crossing =

Correct Classification



Crossing =

Incorrect Crossing



Multi-Objective Graph Embedding

Add classifier for each potential edge/node crossing
parameterized by U

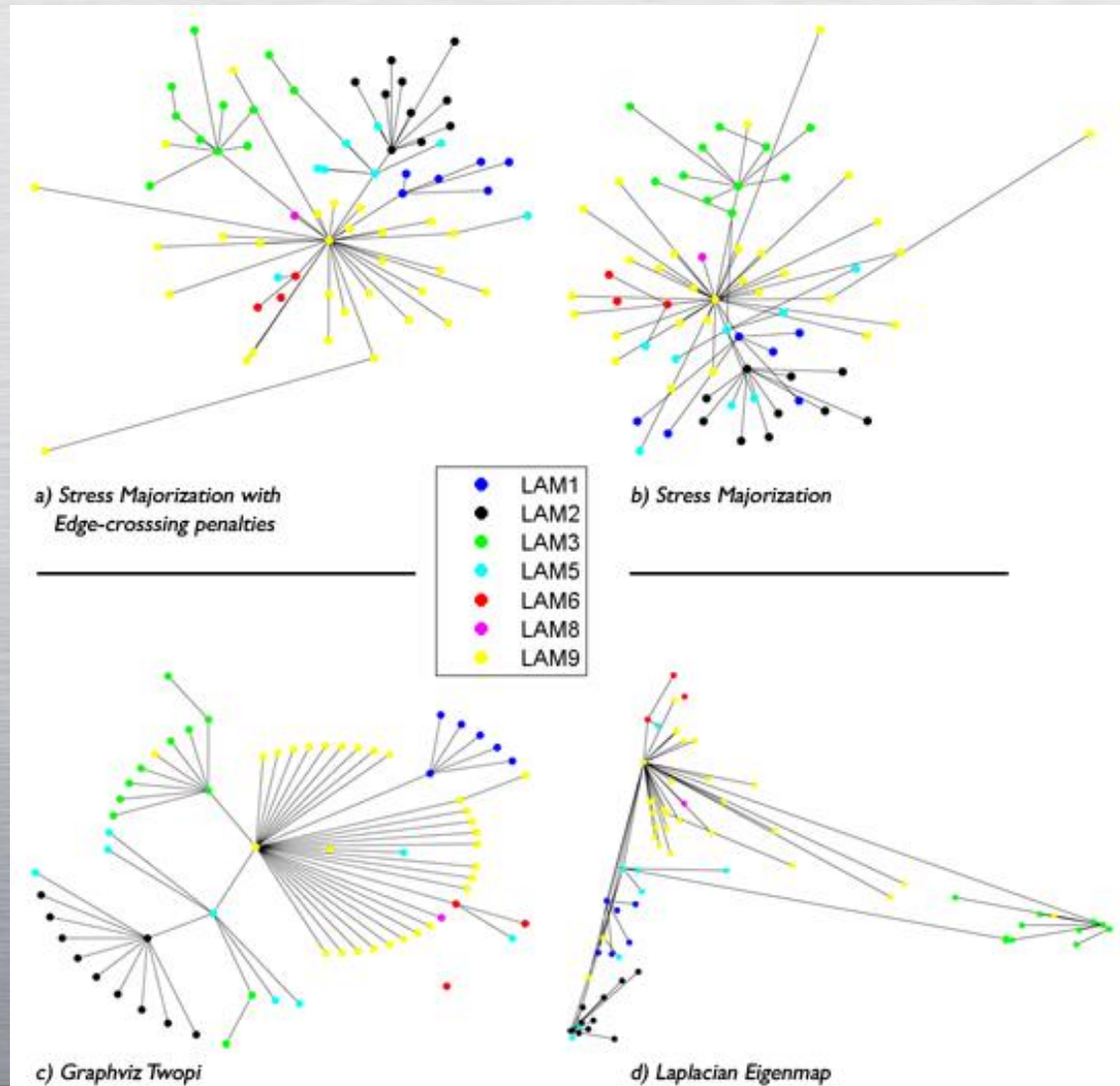
embedding error on X (MDS stress)

+ misclassification error based on U (SVM max
margin)

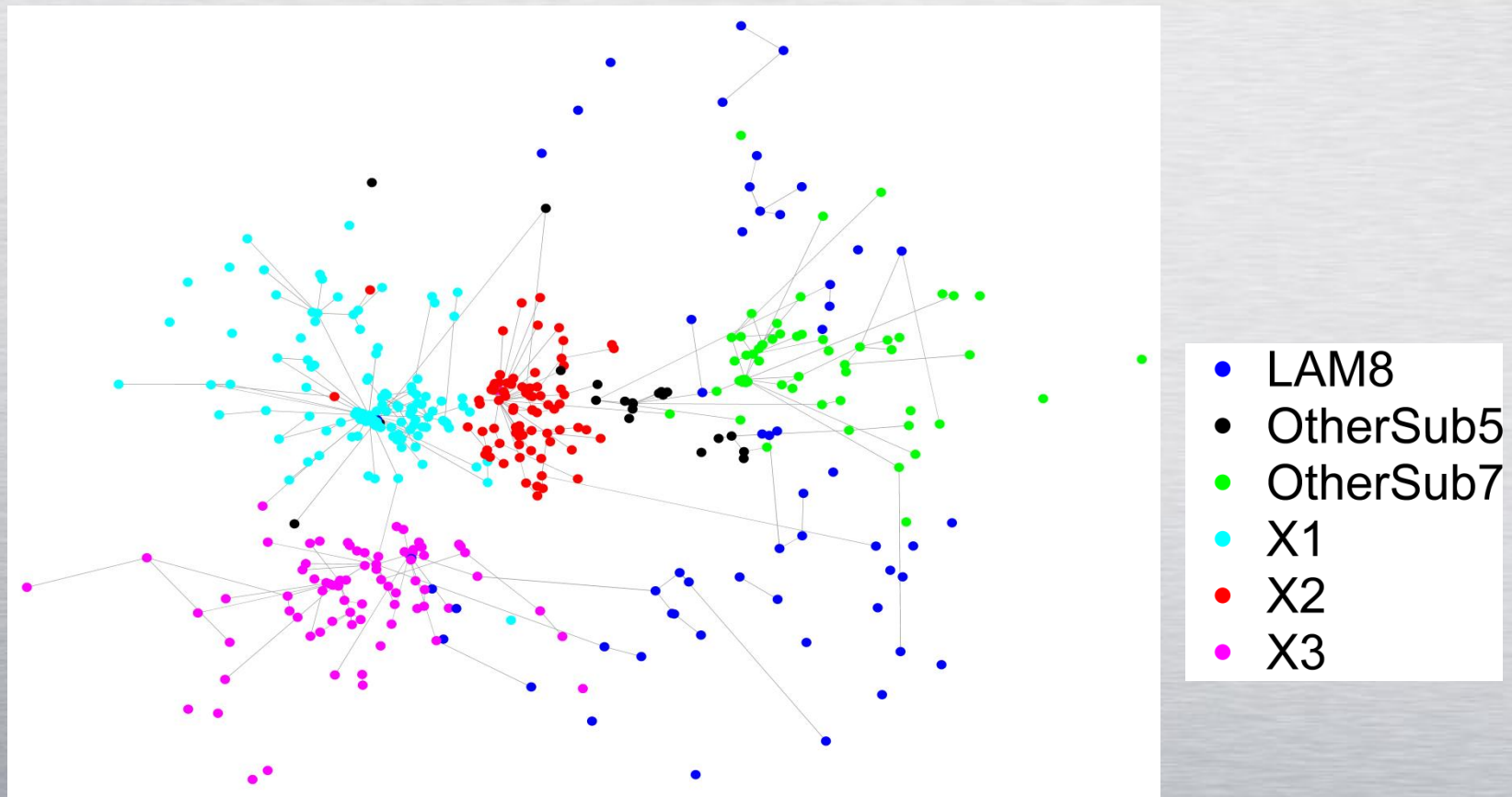
$$\min_{X,u} \text{Stress}(X) + \sum_{i=1}^m \rho_i [\| (-A_i(X)u_i + 1)_+ \|_1 + \| (B_i(X)u_i + 1)_+ \|_1]$$

Optimize by an alternating algorithm on X, U using
scalable classification and embedding algorithms

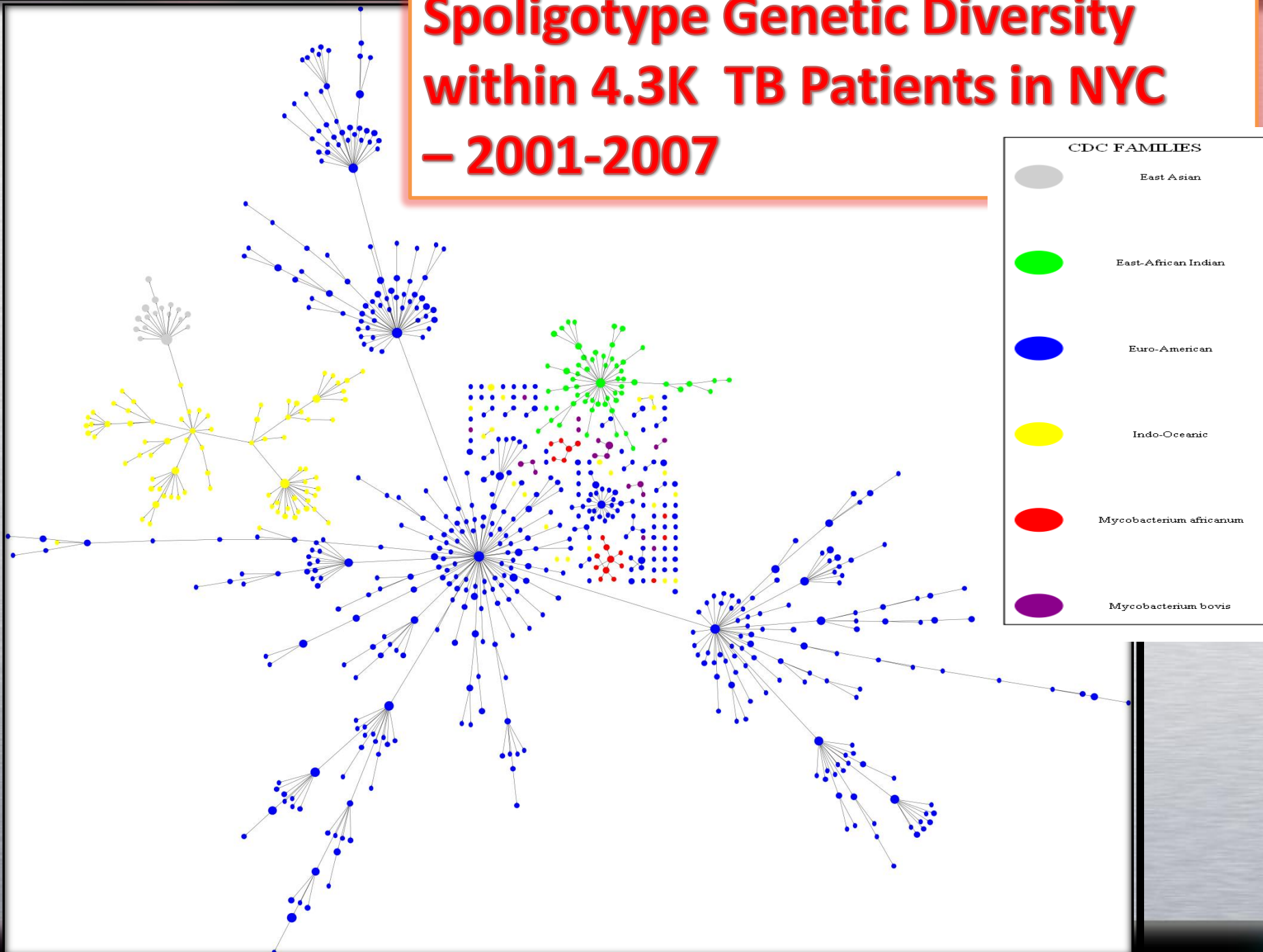
Comparison: LAM sublineages



Visualizing X Lineage adding MIRU and Spoligotype distances



Spoligotype Genetic Diversity within 4.3K TB Patients in NYC – 2001-2007



Host-Pathogen Graphs

Patients = Circles

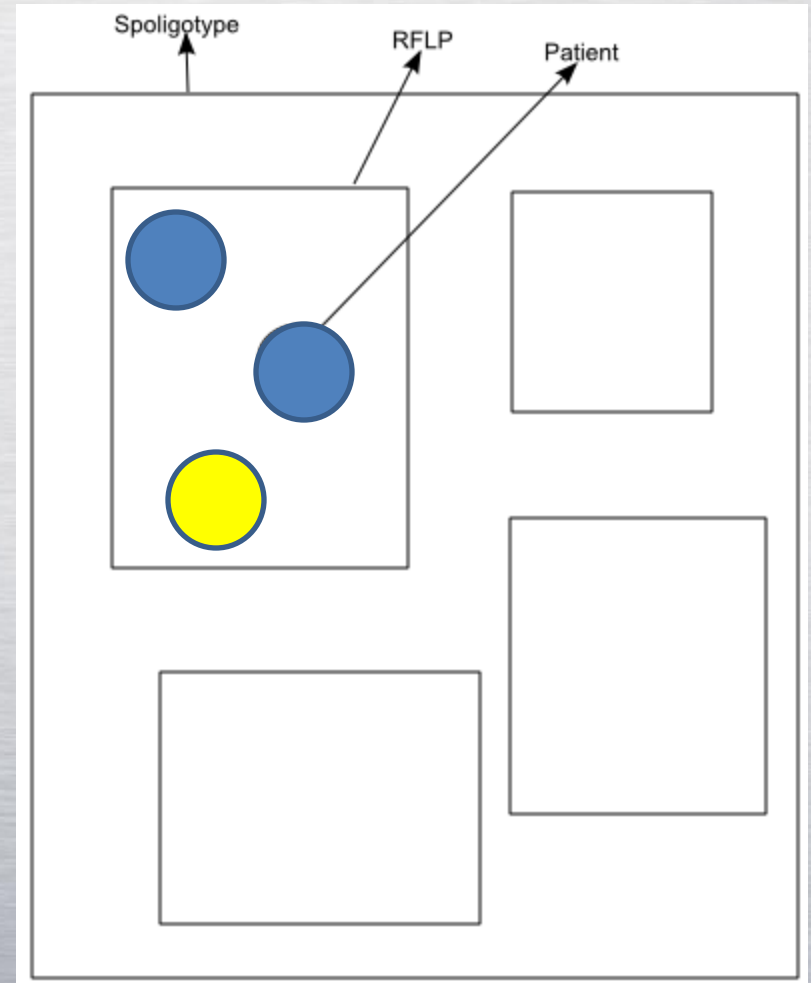
DNA Fingerprint = Box

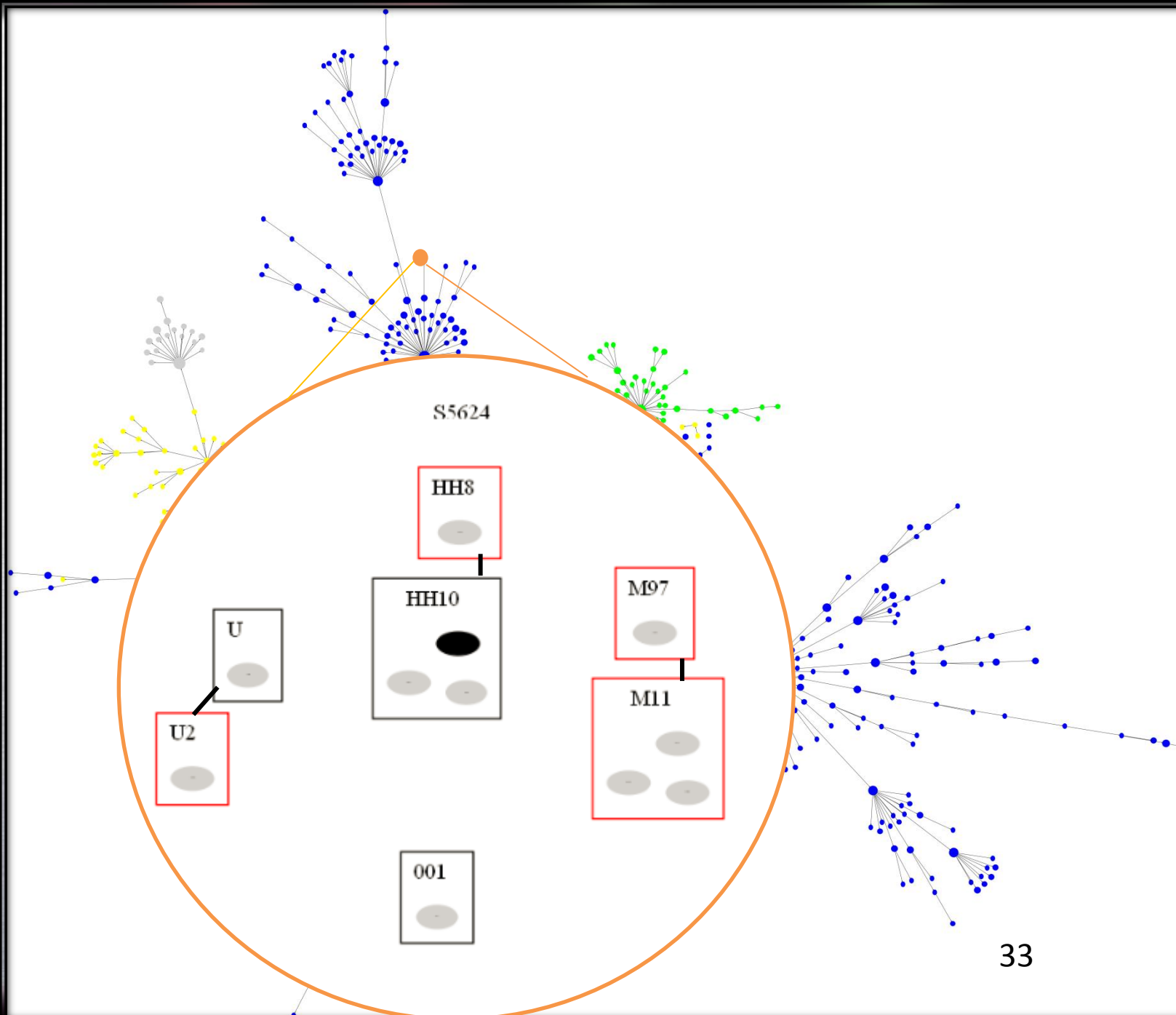
Boxes nested to indicate multiple DNA fingerprints

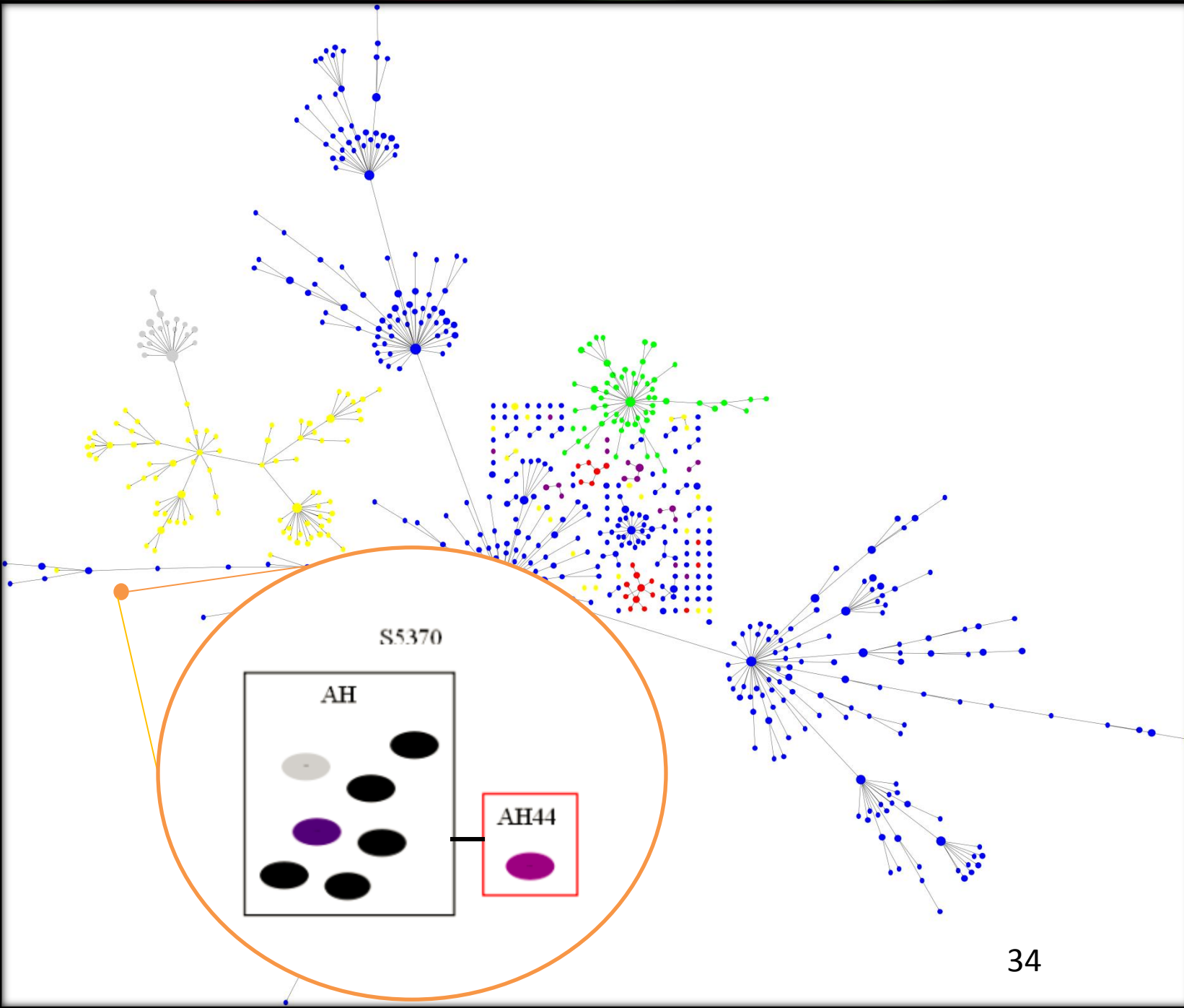
- Other box = spoligotypes
- Inner box = RFLP.

Color by Patient property =
Region of Birth

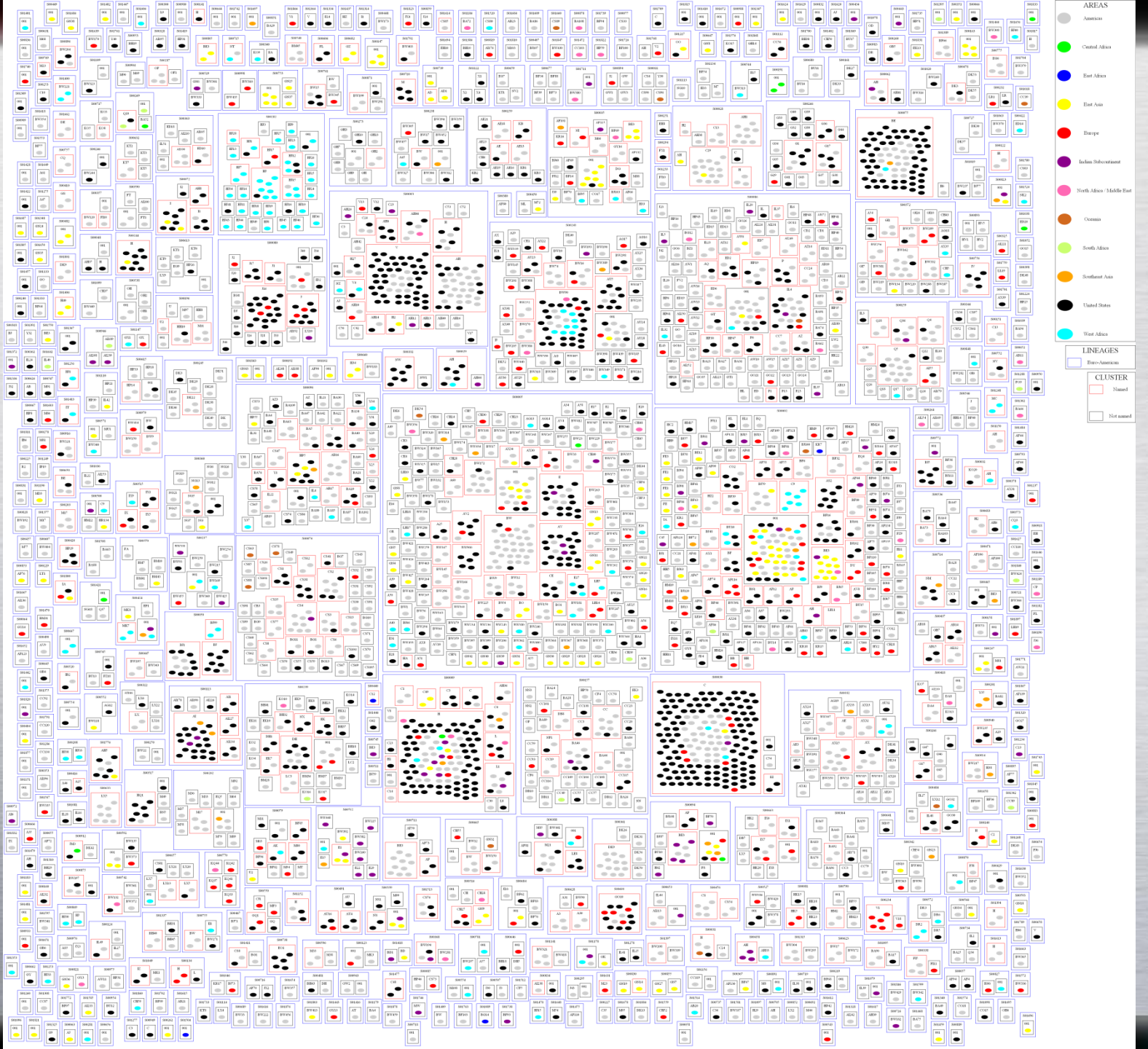
Split by lineage







Euro American



Disease as Stock Market

Companies=Bacteria



Google



BOEING



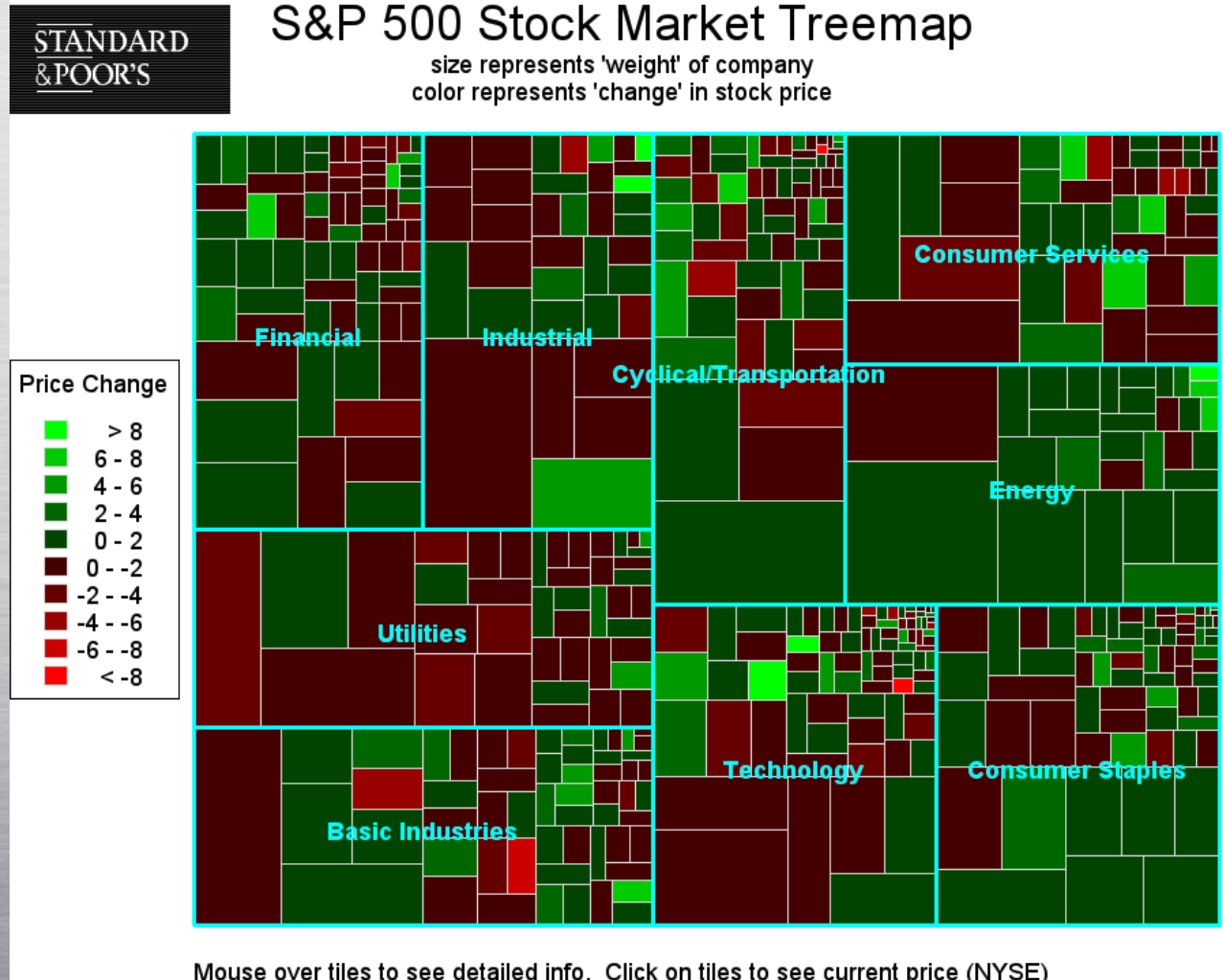
CDC



Buyers = Patients

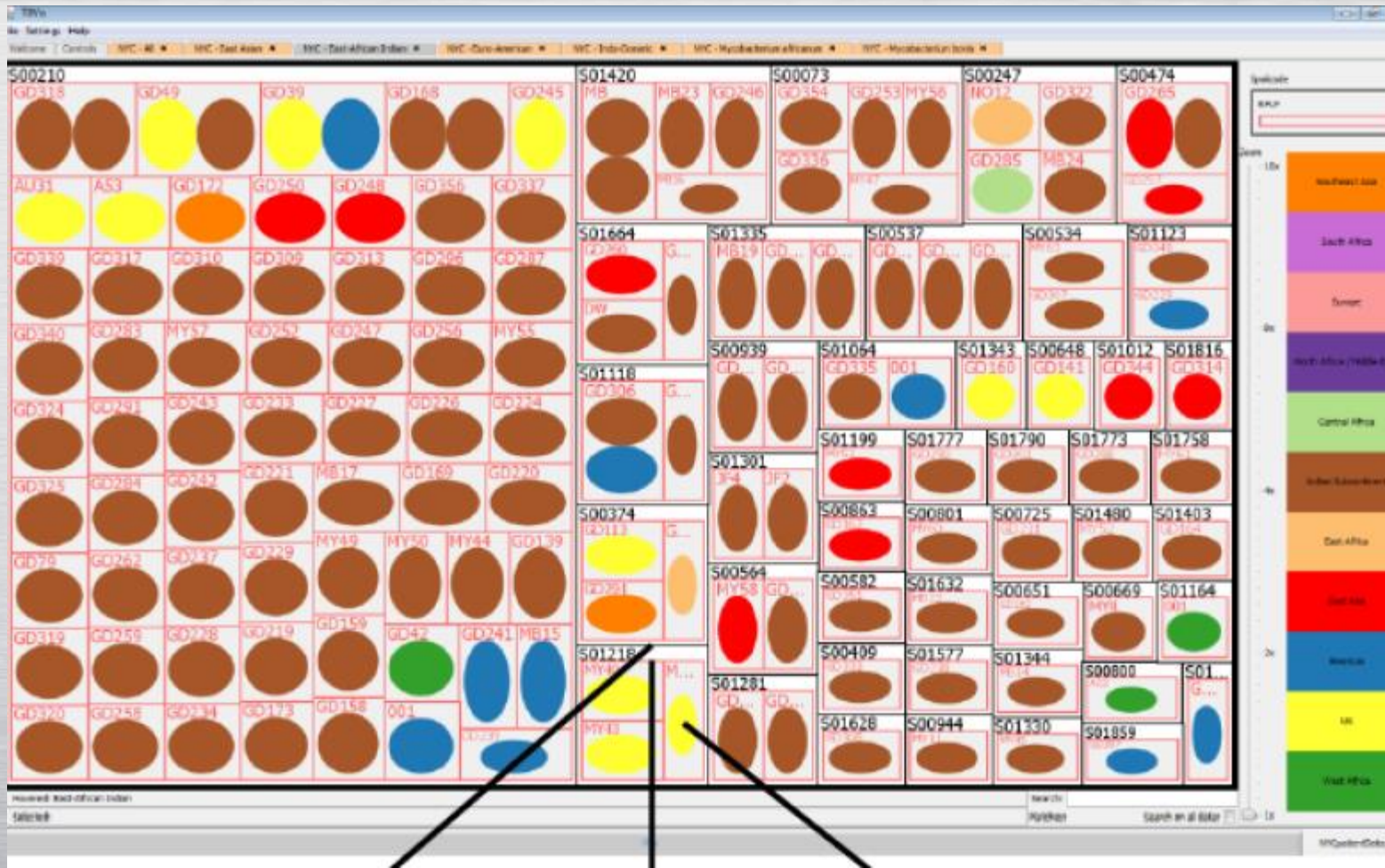


Stock Market Tree Map



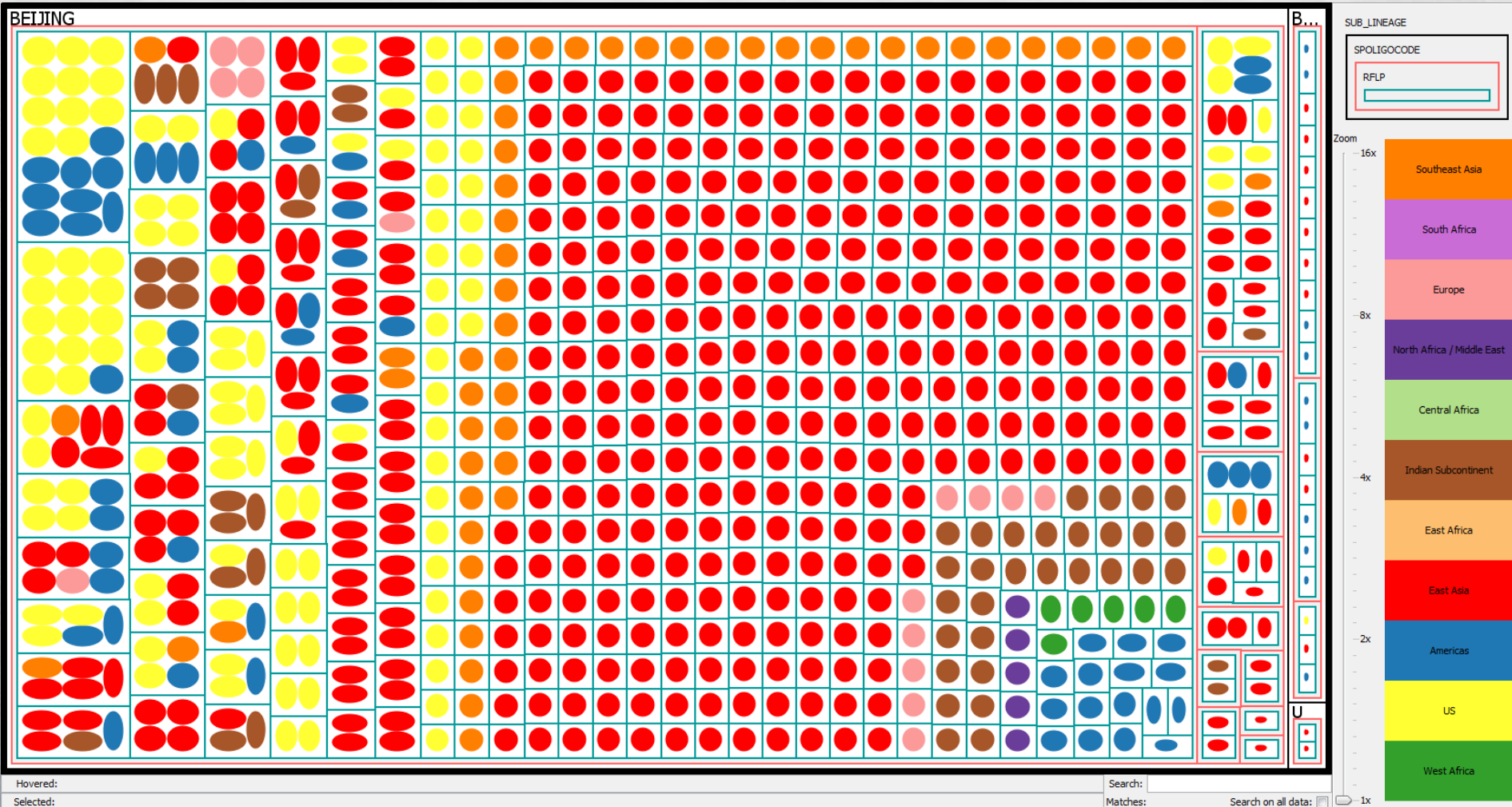
<http://robslink.com/SAS/democd9/sp500.htm>

Split by Lineage - East African Indian

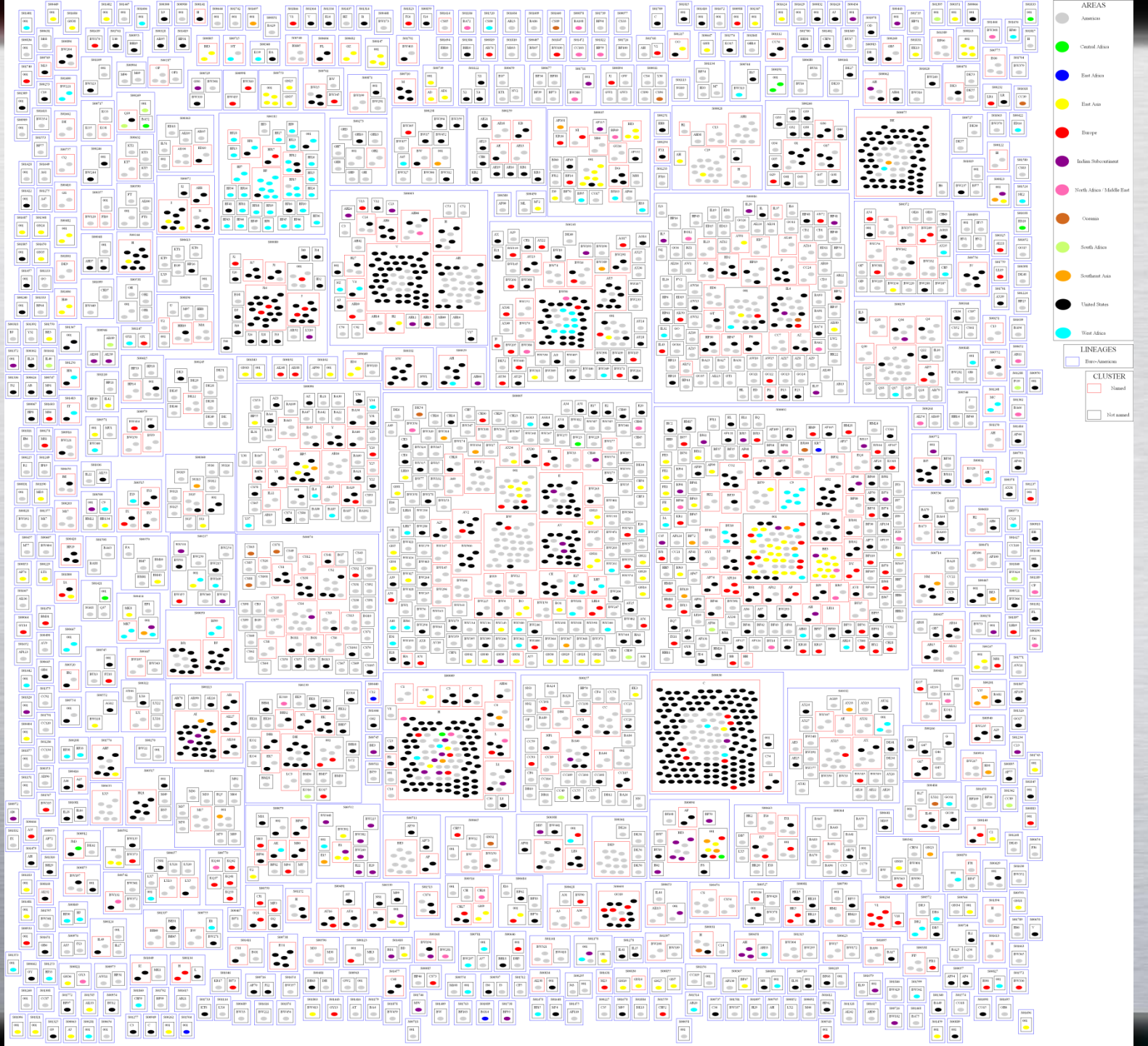


Patient	ID	Biomarker 1	Biomarker 2	TB continent
Patient 1	105	S00669	MY8	Indian Subcontinent
Patient 2	2443	S00210	GD139	Indian Subcontinent
Patient 3	2452	S00210	MY44	Indian Subcontinent
Patient 4	2487	S00247	NO12	East Africa

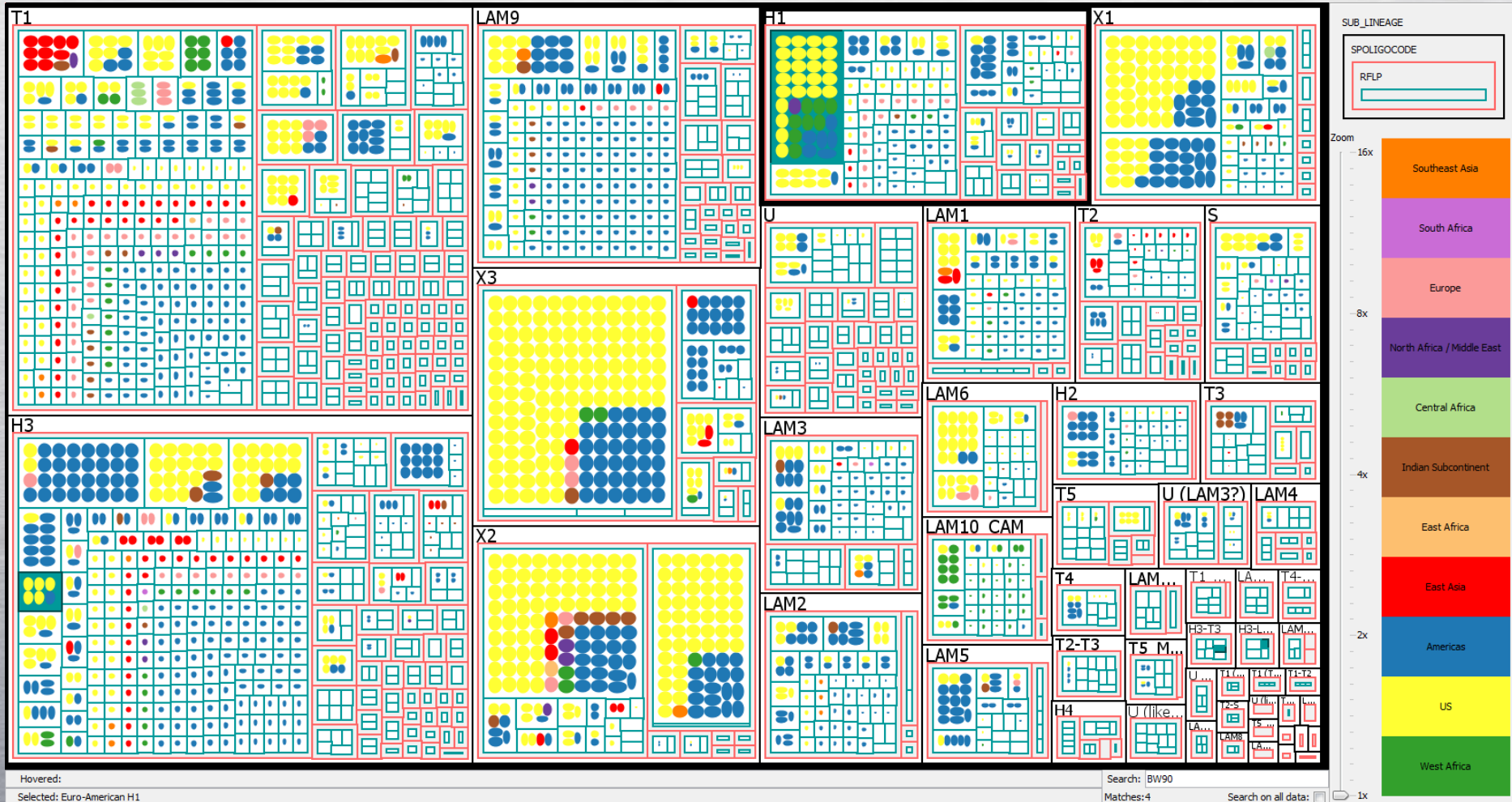
NYC - East Asian



Euro American

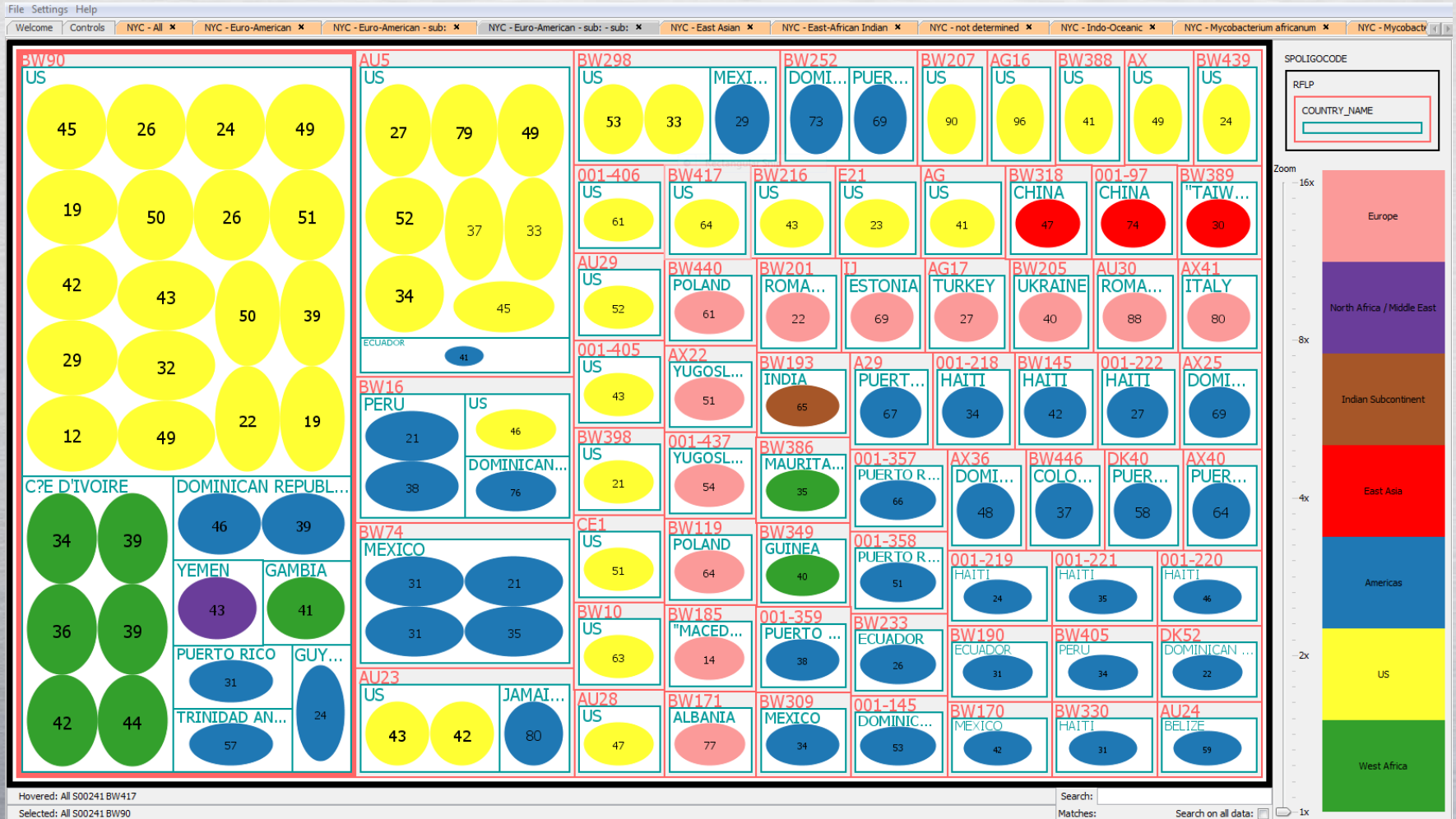


NYC - Euro-American



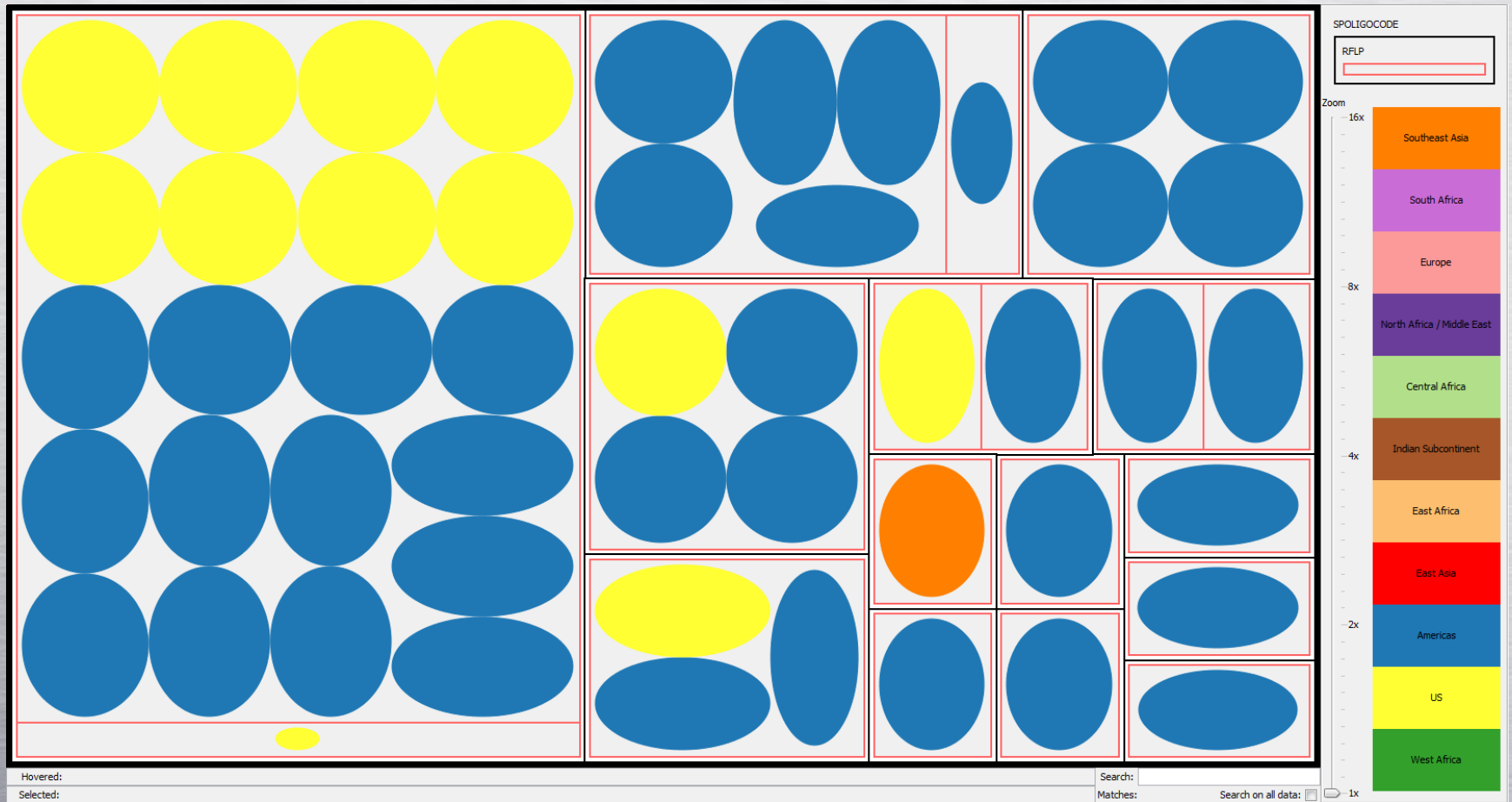
Spoligotype – S00241

BW90 with Patient Age

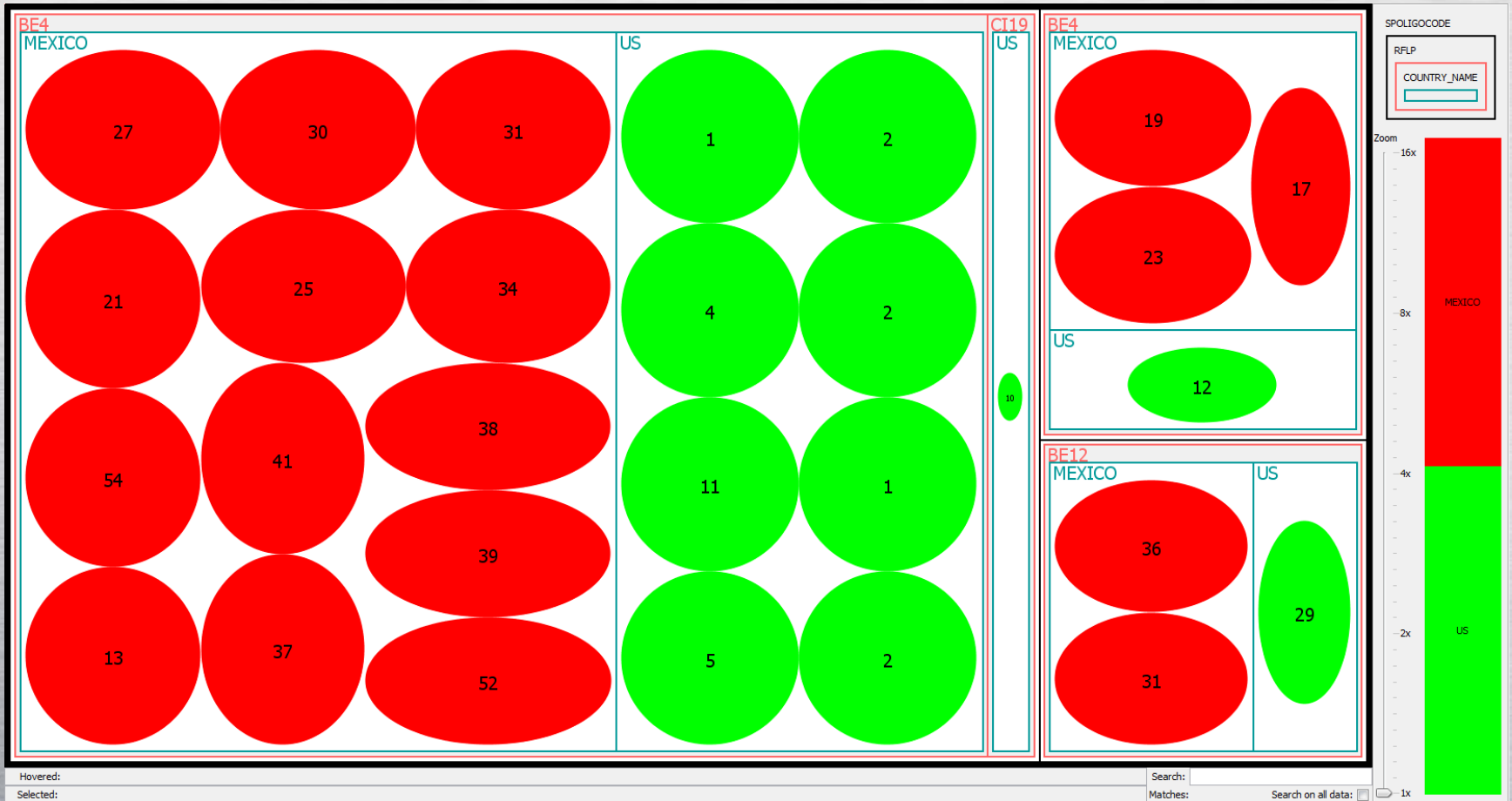


Ongoing transmission in the US that became resistant to INH antibiotic

NYC - *M. bovis*

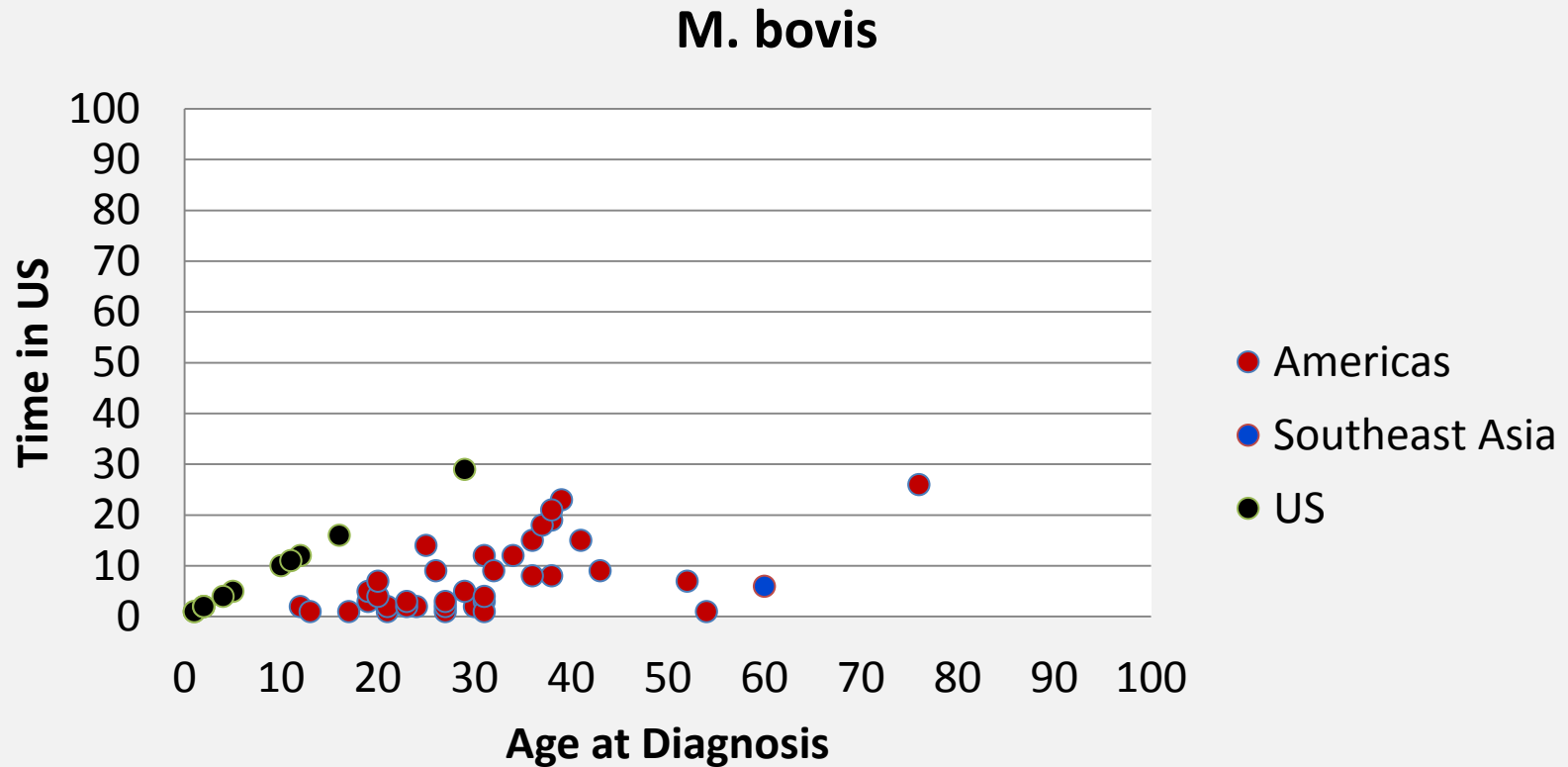


M. Bovis with Age of Patient/Country



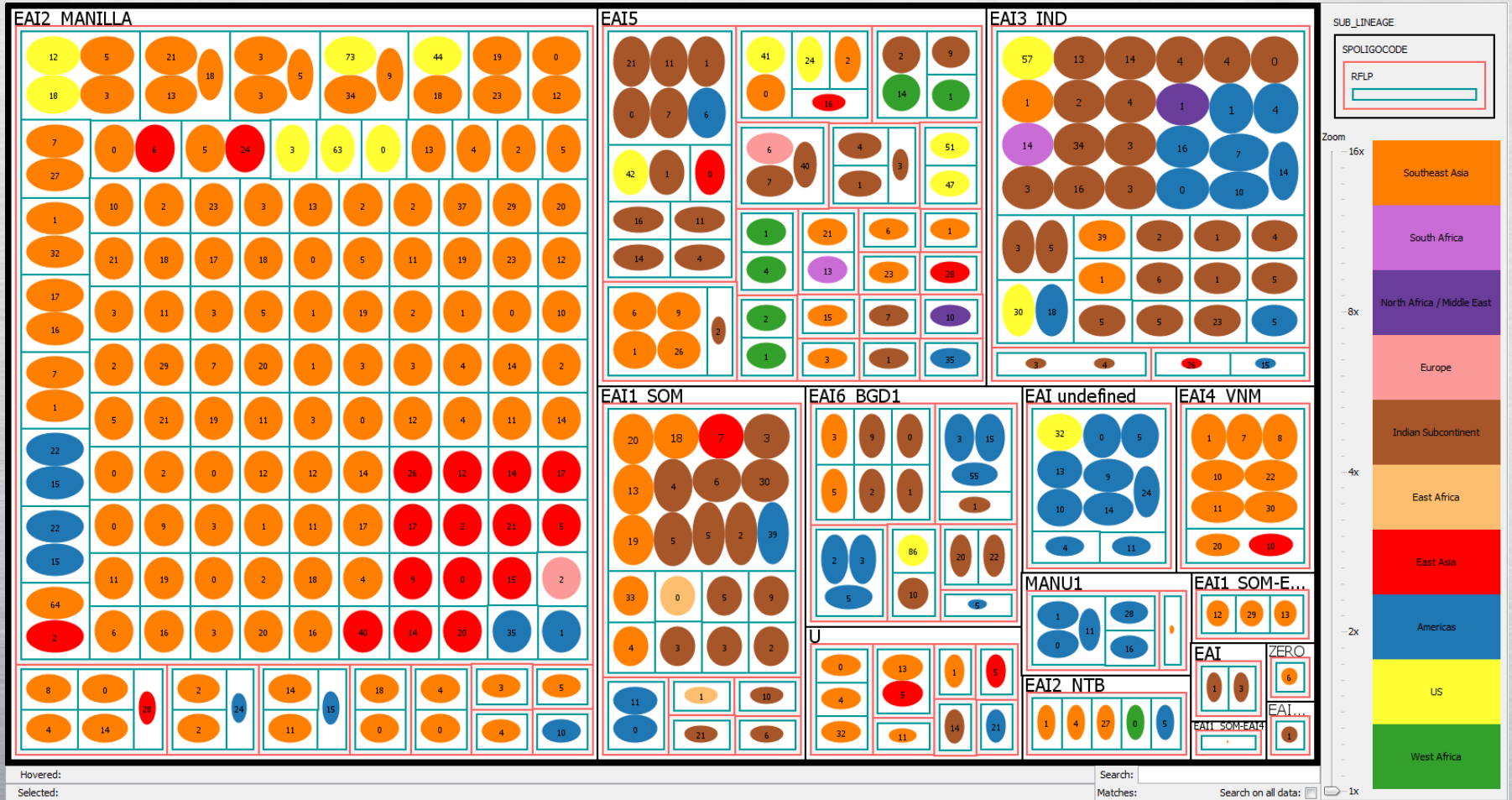
Extremely Young: Outbreak in US-born children of Mexican Parents likely due to unpasteurized cheese.

NYC *M. bovis* (2001-2007)



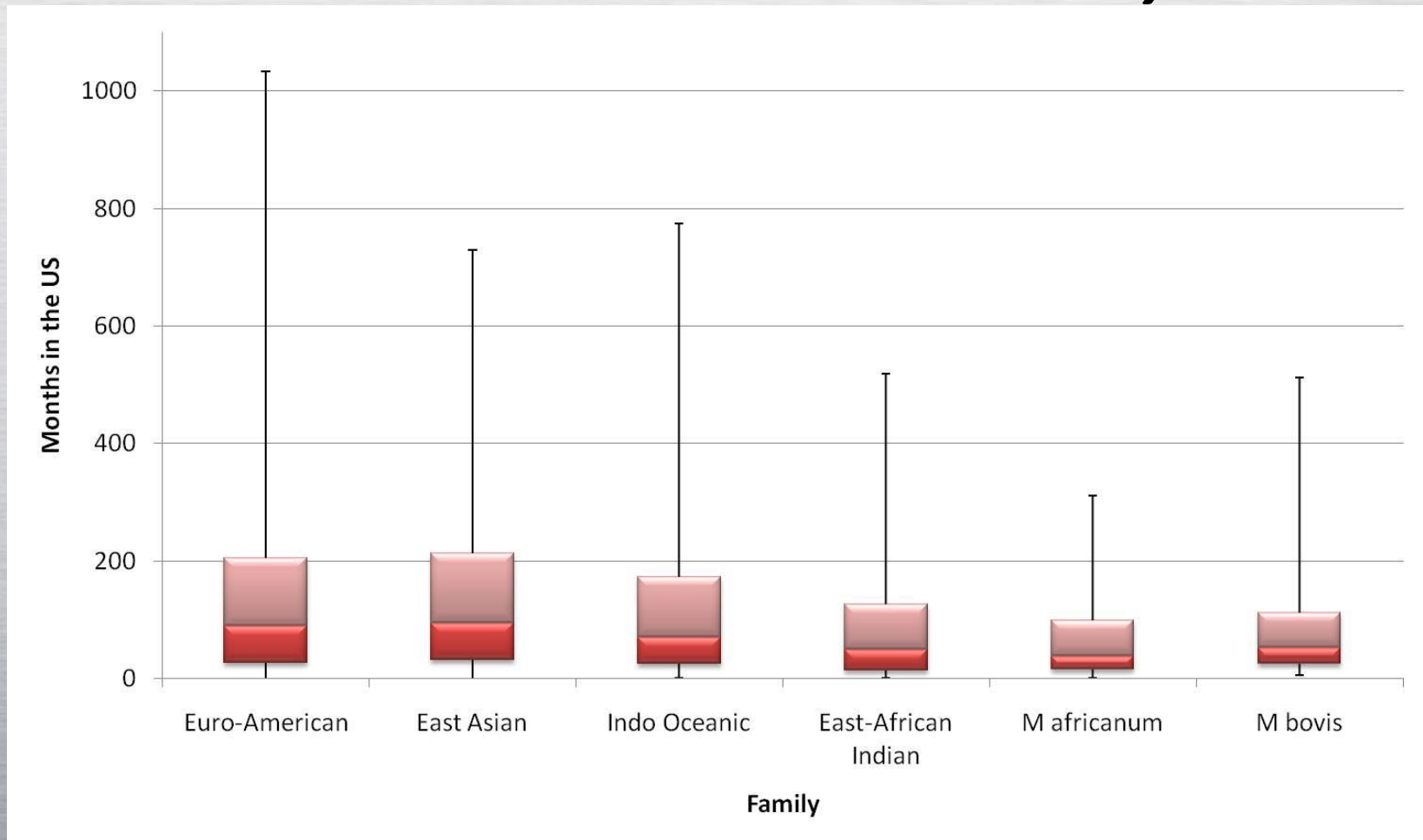
- Extra pulmonary *M. bovis* strikes
 - Mexican Immigrants
 - US-born children of Mexican Immigrants
- Hypothesized caused: Unpasteurized cheese

Indo-Oceanic with Time in US



Large clusters with few US patients and no found epi-links.

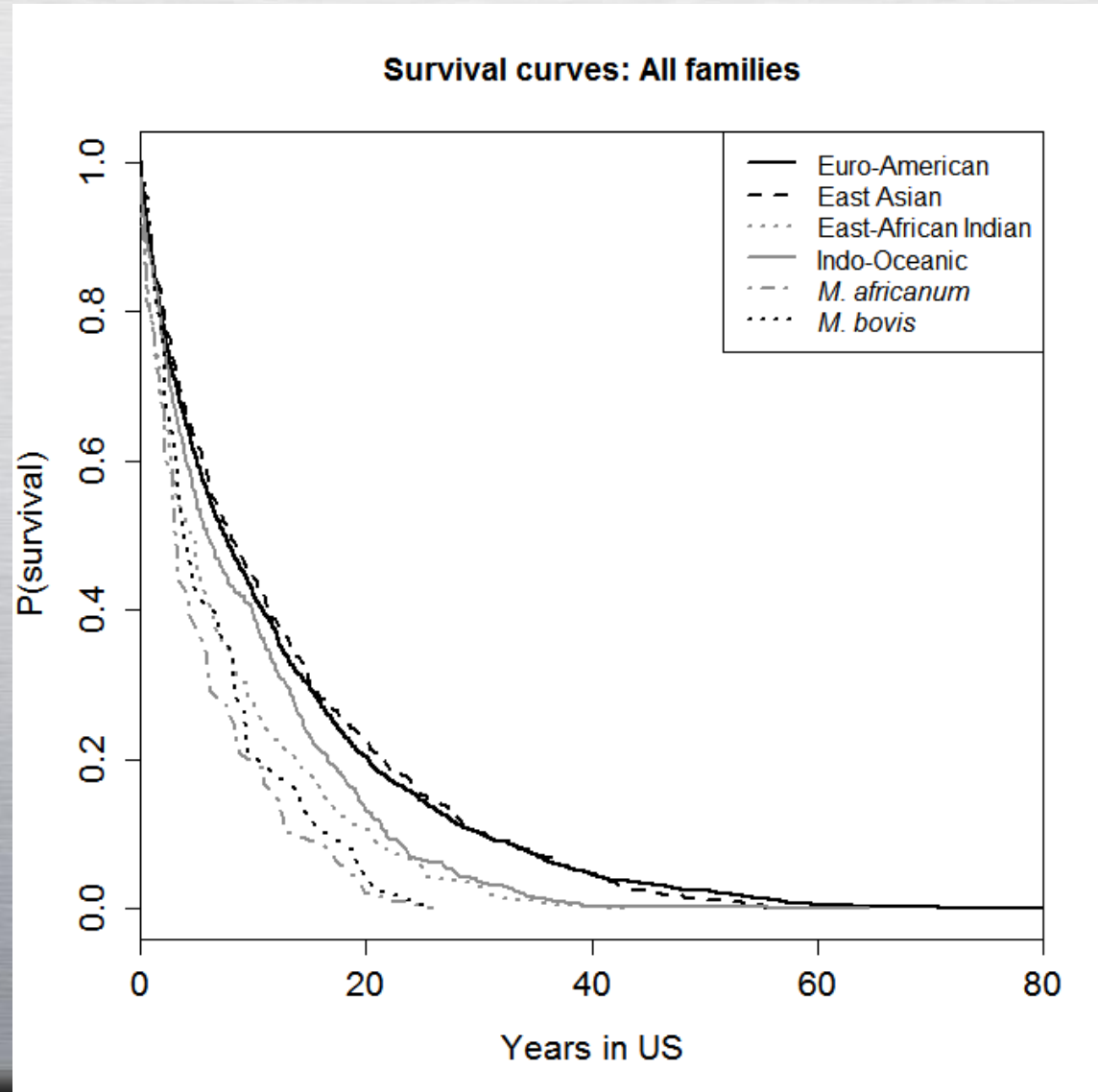
Indo-Oceanic Anomaly



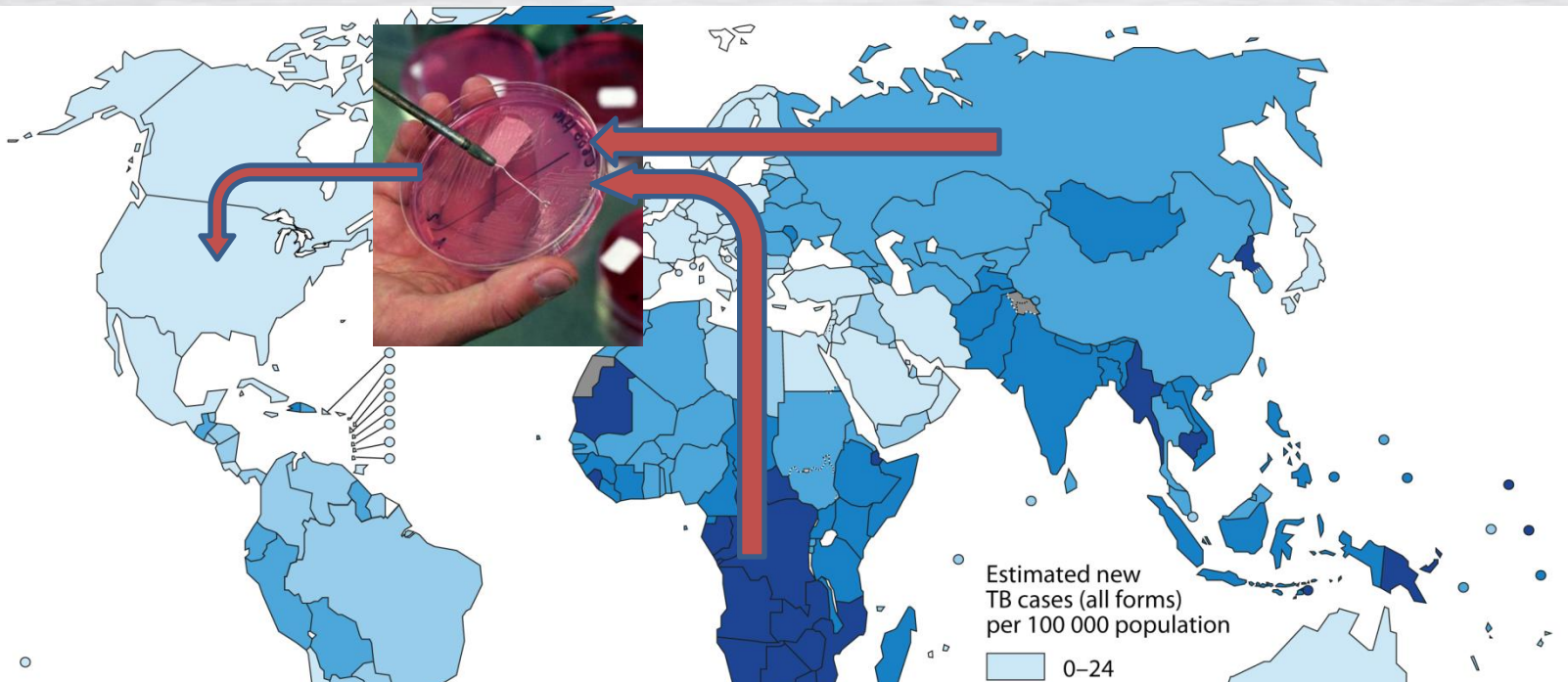
Hypothesized cause: IO strains have longer latency phenotype

Survival curves for all lineages





Proportion of cases not yet activated



Surveillance Data can reveal novel phenotypes and genotypes



Whole Genome Sequencing versus Traditional Genotyping for Investigation of a *Mycobacterium tuberculosis* Outbreak: A Longitudinal Molecular Epidemiological Study

Andreas Roetzer , Roland Diel , Thomas A. Kohl , Christian Rückert, Ulrich Nübel, Jochen Blom, Thierry Wirth, Sebastian Jaenicke, Sieglinde Schuback, Sabine Rüscher-Gedes, Philip Supply, Jörn Kalinowski, Stefan Niemann 

Challenges of Disease Control using Molecular Epidemiology

- **Coupling human intelligence and analytics to help public health care workers control disease**
- **Informing local decisions with global data**
- **Allocating scarce control resources effectively by predicting disease dynamics**
- **Incorporating rapidly evolving data**
 - Contact Investigations
 - New biomarkers for pathogen/host
 - Electronic Medical Records
 - Social media
- **Getting the biology right**
- **Preserving Privacy**



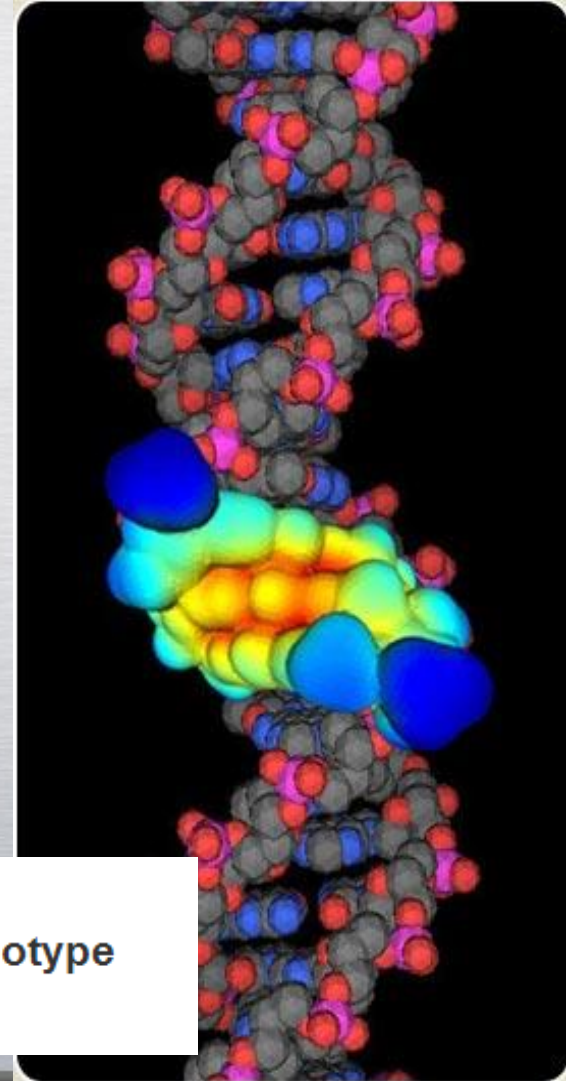
Personalized Medicine based on Host and Pathogen DNA



- Discovery of Host/Pathogen Coadaptation
- Control and treatment efforts guided by host and pathogen DNA
- Better models for drug development, etc

Ethnicity and mycobacterial lineage as determinants of tuberculosis disease phenotype

Thorax, 2013

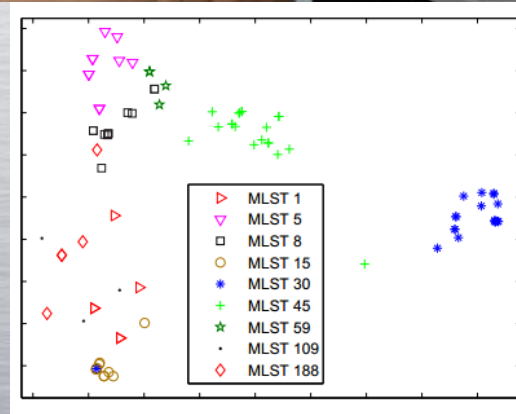
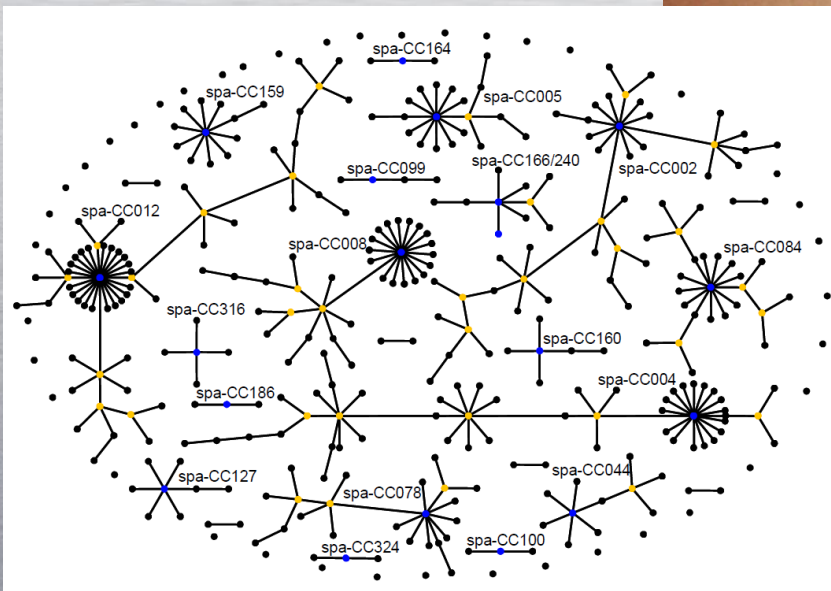


Other Diseases/Pathogens

MRSA



Dr. Gregory Moran, M.D.



DATA ANALYSIS

DATABASES

- ▣ *B.burgdorferi*
- ▣ *B.cereus*
- ▣ *B.henselae*
- ▣ *B.pseudomallei*
- ▣ *C.albicans*
- ▣ *C.glabrata*
- ▣ *C.trachomatis*
- ▣ *C.krusei*
- ▣ *C.tropicalis*
- ▣ *C.jejuni*
- ▣ *C.neoformans var grubii*
- ▣ *E.coli*
- ▣ *E.faecalis*
- ▣ *E.faecium*
- ▣ *H.influenzae*
- ▣ *H.pylori*
- ▣ *Leptospira spp.*
- ▣ *M.catarrhalis*
- ▣ *N.meningitidis*
- ▣ *P.acnes*
- ▣ *S.agalactiae*
- ▣ *S.aureus*
- ▣ *S.dysgalactiae*
- ▣ *S.enterica*
- ▣ *S.epidermidis*
- ▣ *S.pneumoniae*
- ▣ *S.pyogenes*
- ▣ *S.suis*
- ▣ *Yersinia*

TB-Insight Project Team

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**During this talk approximately
998 people developed active TB
276 people died of TB**

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