

Advancing Forensic DNA Analysis in Support of DoD Missions



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Advancing Forensic DNA Analysis in Support of DoD Missions

Ellen Greytak, PhD

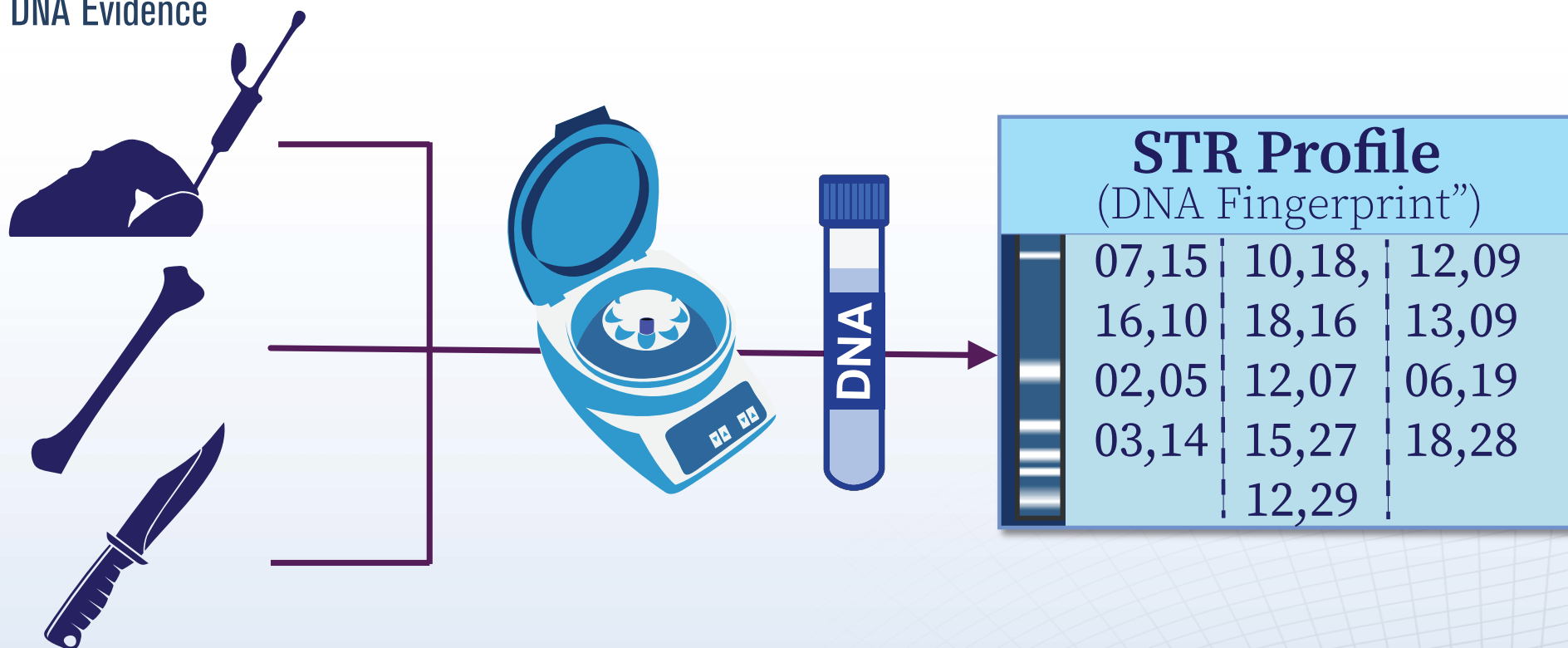
Director of Bioinformatics

Parabon NanoLabs, Inc.



Traditional Forensic DNA Analysis

DNA Evidence

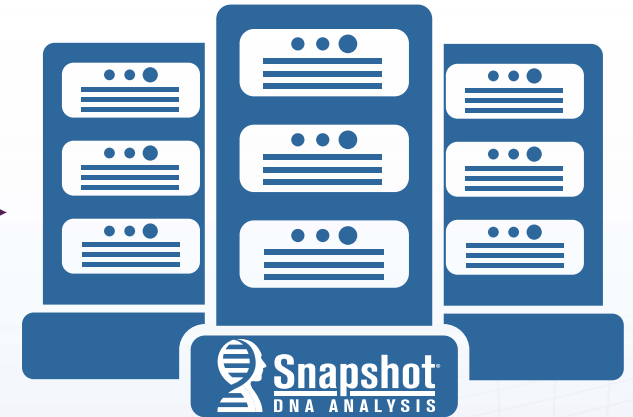


- › Short Tandem Repeats (STRs)
- › Identity questions ONLY
 - Is this person in a database (e.g., CODIS)?
 - Is this person a suspect we've already identified?
 - Definitively establishes identity

Advanced Forensic DNA Analysis



SNP Genotype ("DNA Blueprint")					
CC	CC	CT	CC	GG	CC
GG	AA	AA	AG	CC	CC
CC	AA	AA	TT	CC	TT
CT	AA	TT	AA	CC	AG
AA	CT	AG	CC	CC	AA



- › Single Nucleotide Polymorphisms (SNPs)
- › Generate NEW leads
 - What does this unknown person look like?
 - Is this unknown person related to someone we DO know? If so, how closely?
 - Is this unknown person related to volunteers in genetic genealogy databases?
 - Generates leads, but identity is established through traditional methods

DoD-Funded Projects

DNA Phenotyping

- DTRA Phase I and Phase II SBIRs
- Sequential Phase II SBIR

Distant Kinship Analysis

- DTRA SBIR Modification
- ARO Contract

Genetic Genealogy

- Direct result of Phenotyping and Kinship projects

Next-Generation Sequencing (NGS) Analysis

- ARO Contract

DoD-Funded Project Success

DNA Phenotyping

- Commercialized for law enforcement

Distant Kinship Analysis

- Commercialized for law enforcement
- Transitioning to AFDIL

Genetic Genealogy

- Commercialized for law enforcement
- 111 solved cases (and counting!)

Next-Generation Sequencing (NGS) Analysis

- Transitioned to DoD forensics labs

DNA Phenotyping

Snapshot®: Predict the physical appearance of an unknown person from DNA

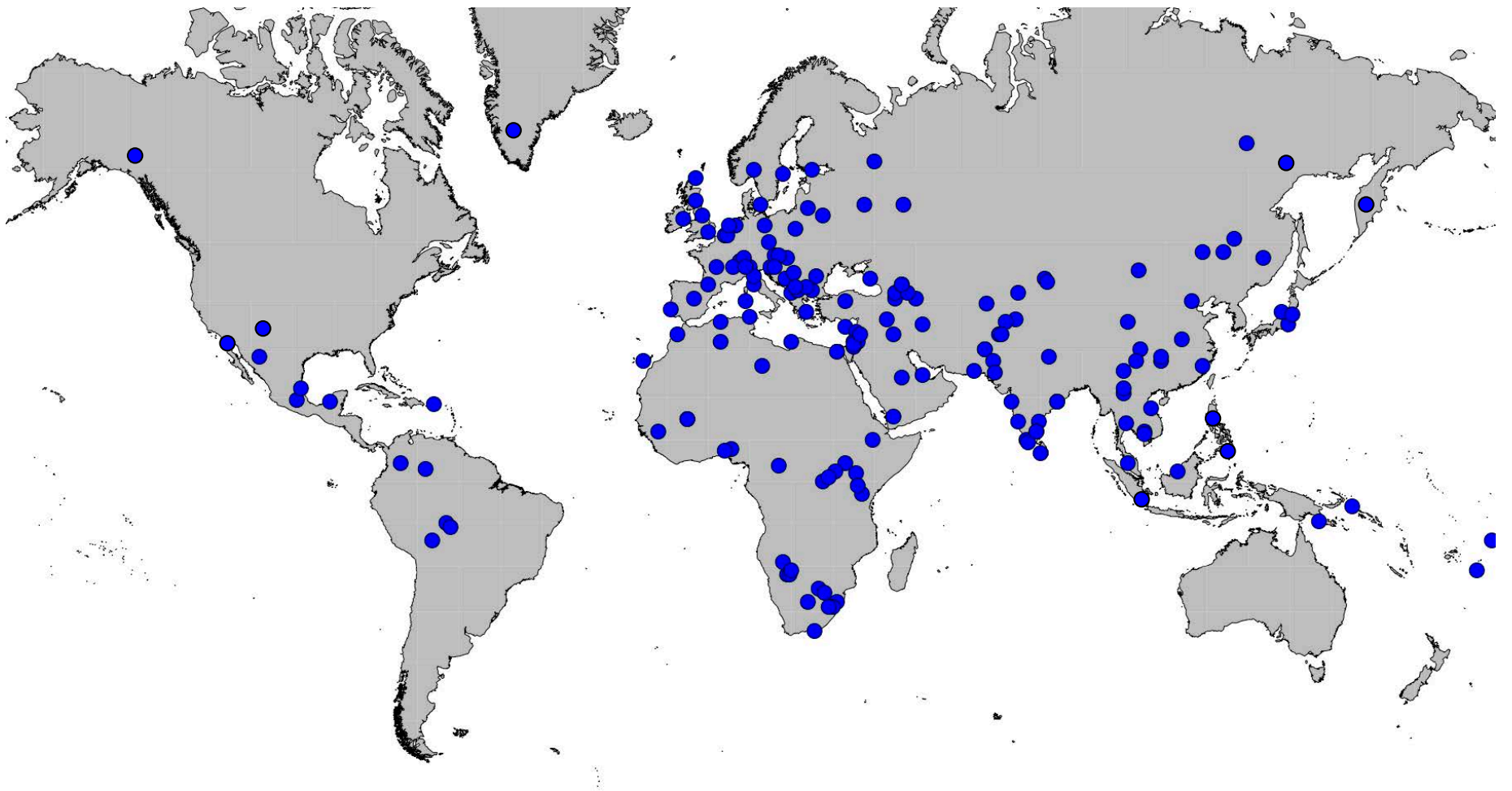


DNA Phenotyping

- › **DoD Problem:** in theaters of war, a DNA database may not be available for traditional forensic DNA comparisons
- › **DTRA SBIR solicitation goal:** given a DNA sample from an unexploded IED, predict the appearance of the person who planted it

Ancestry Inference

- >13,000 background subjects from >150 populations (aggregated from multiple studies for worldwide coverage)

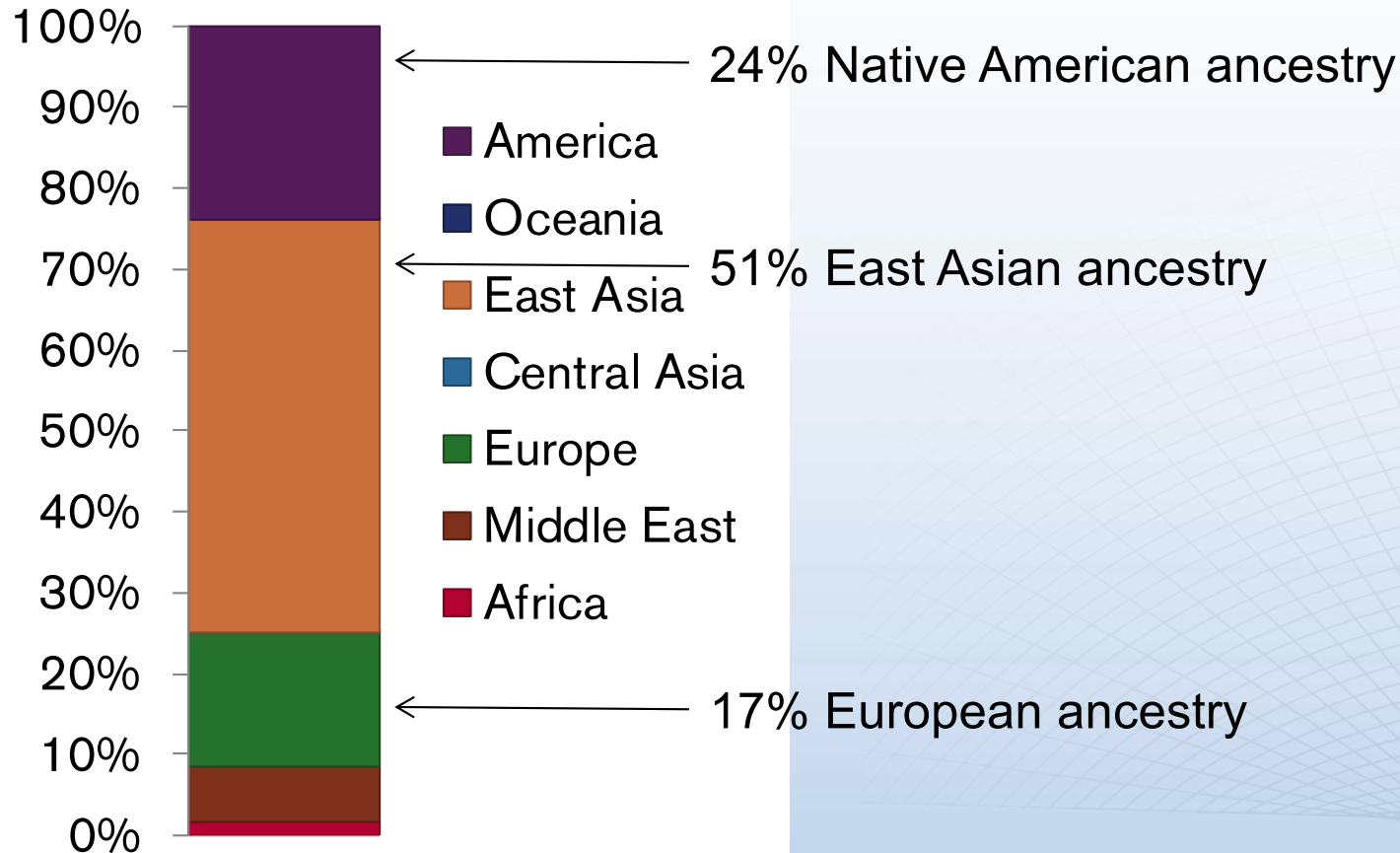


Ancestry Inference

- › Tens of thousands of SNPs across the genome
- › Very precise estimates of ancestry across as many populations as are defined in the background
- › Can detect even low levels of admixture (even when the mixture is between populations we haven't previously observed)

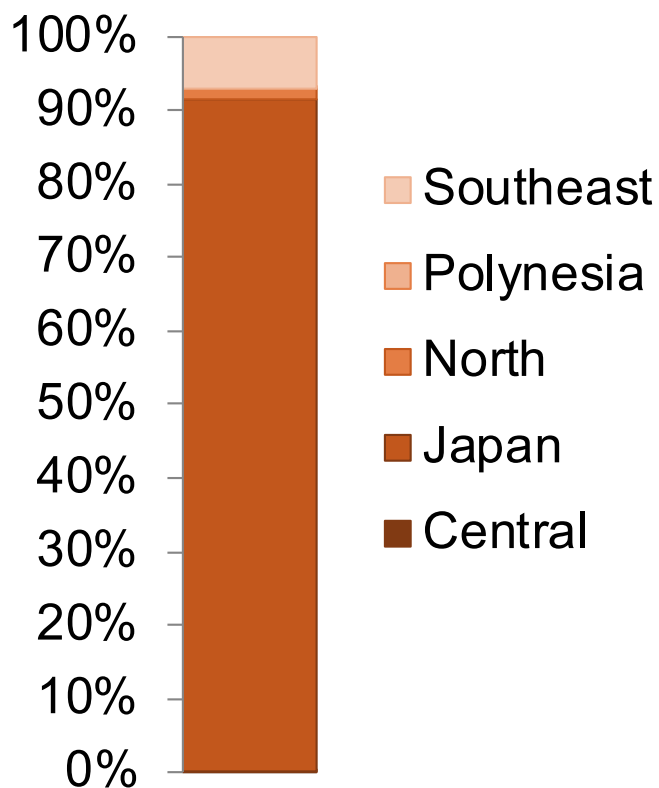
Global Ancestry: Admixed Individual

- Real results from an unknown individual given to us during a blind evaluation test

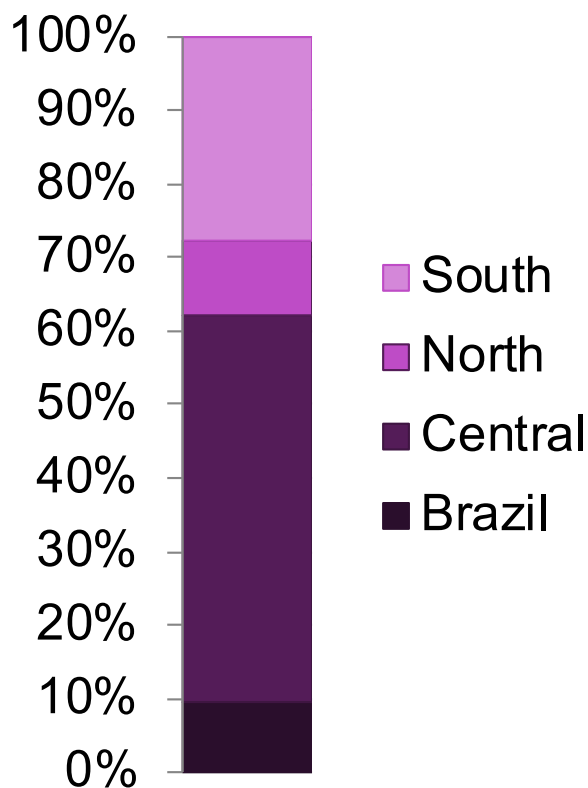


Regional Ancestry: Admixed Individual

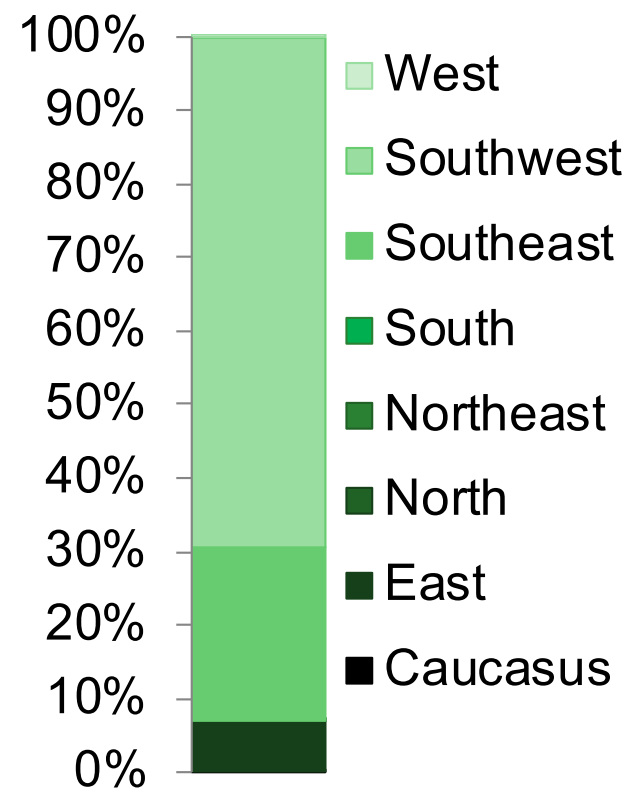
East Asia



America



Europe



- › Conclusion: This individual is half Japanese and half Latino
- › Y chromosome haplogroup = O1b2a (most common in Japan and Korea)
- › Therefore, this individual most likely has a Japanese father and a Central American Latino mother

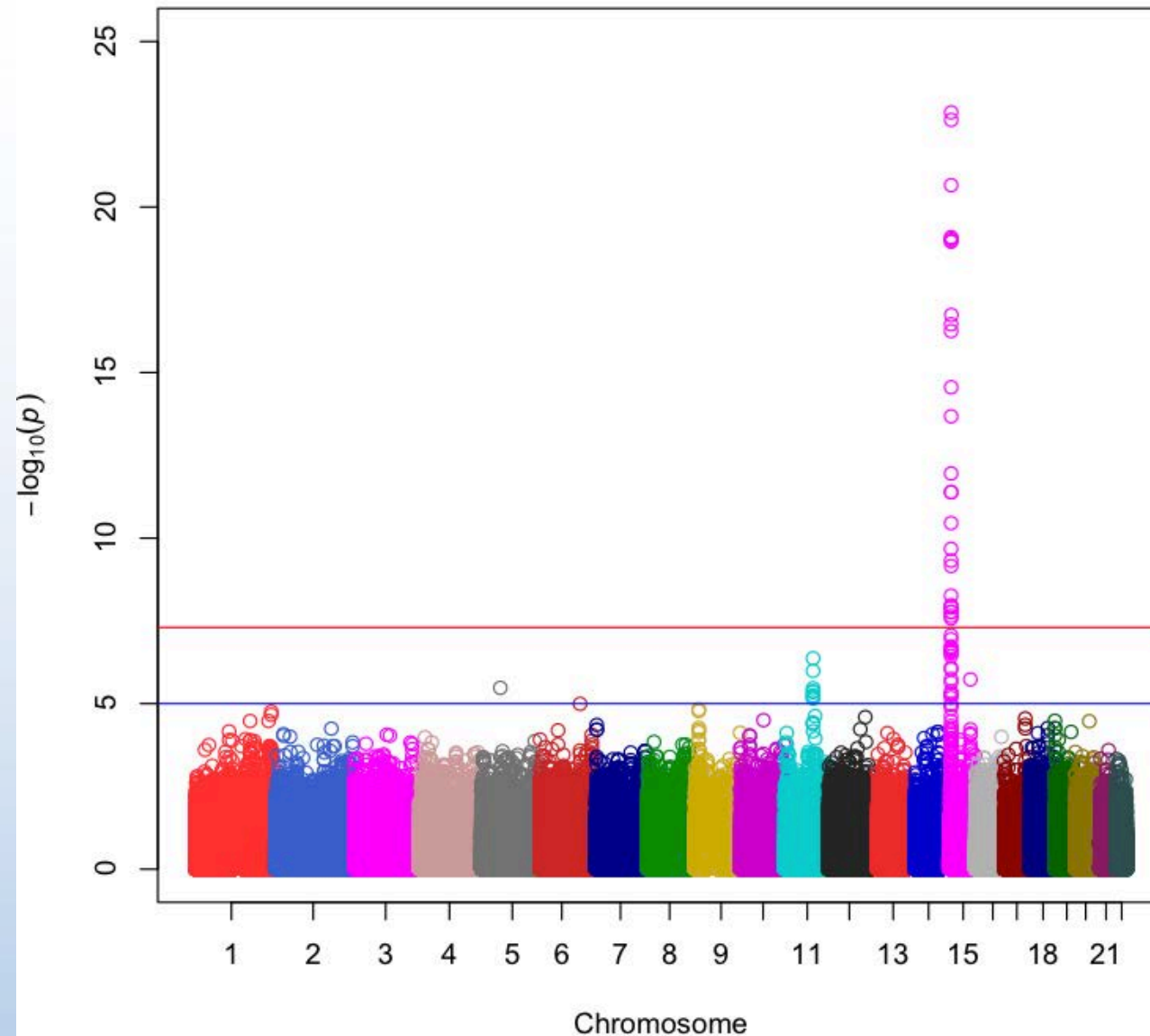
Genome-Wide Association

- › Detect statistical associations between SNPs and a heritable phenotype
- › Requires a large genotype+phenotype (G+P) database of subjects with known phenotype and genome-wide data

Subject	Eye Color	SNP1 Genotype	SNP2 Genotype	SNP3 Genotype	SNP4 Genotype	...	SNP1,000,000 Genotype
1	Blue	A/G	C/C	G/G	G/G	...	T/T
2	Green	A/A	T/T	A/G	A/G	...	T/T
3	Hazel	G/G	C/T	A/G	A/A	...	T/T
4	Brown	A/A	T/T	A/A	G/G	...	C/T
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
3,000	Blue	G/G	T/T	G/G	G/G	...	C/T

Genome-Wide Association

- > A strong statistical association suggests the SNP might be biologically involved in the phenotype

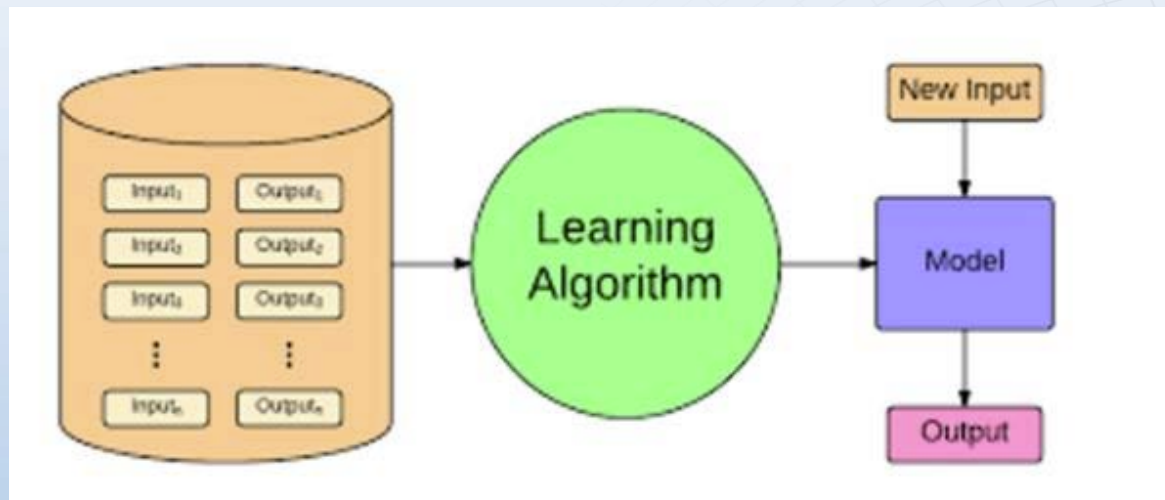


Interaction Analysis

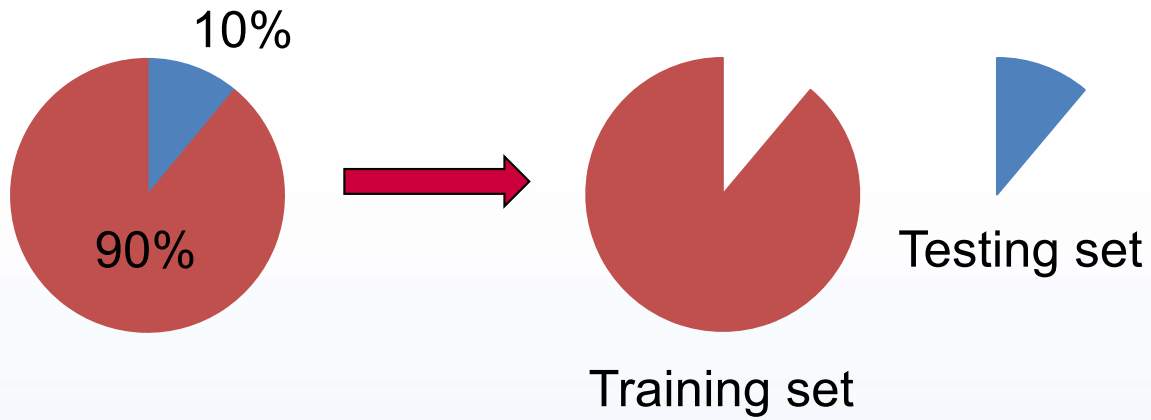
- › Single SNP association testing may not capture the whole story
- › Many traits are influenced not only by individual SNPs but by non-additive (epistatic) *interactions* among several SNPs
- › However, looking for high-order interactions (e.g., 3, 4, and 5 factors) on a genome-wide scale results in a combinatorial explosion of possible models
 - 8,333,250,000,000,000,000,000,000,000 (10²⁷) possible 5-way interactions among 1 million SNPs
- › Parabon has developed software (Crush-MDR) that uses a distributed evolutionary search algorithm to explore the massive space of possible interactions

Genotype-to-Phenotype (G2P) Modeling

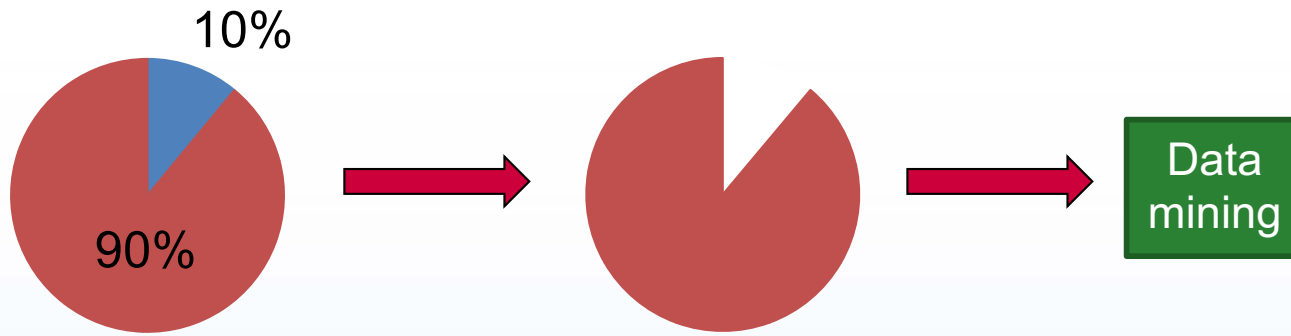
- › Supervised machine learning (ML)
- › Train the model using:
 - Known phenotype values
 - Extracted features (selected SNPs, covariates such as ancestry and sex)
- › Learn a function to predict outcome from features
- › Evaluate accuracy



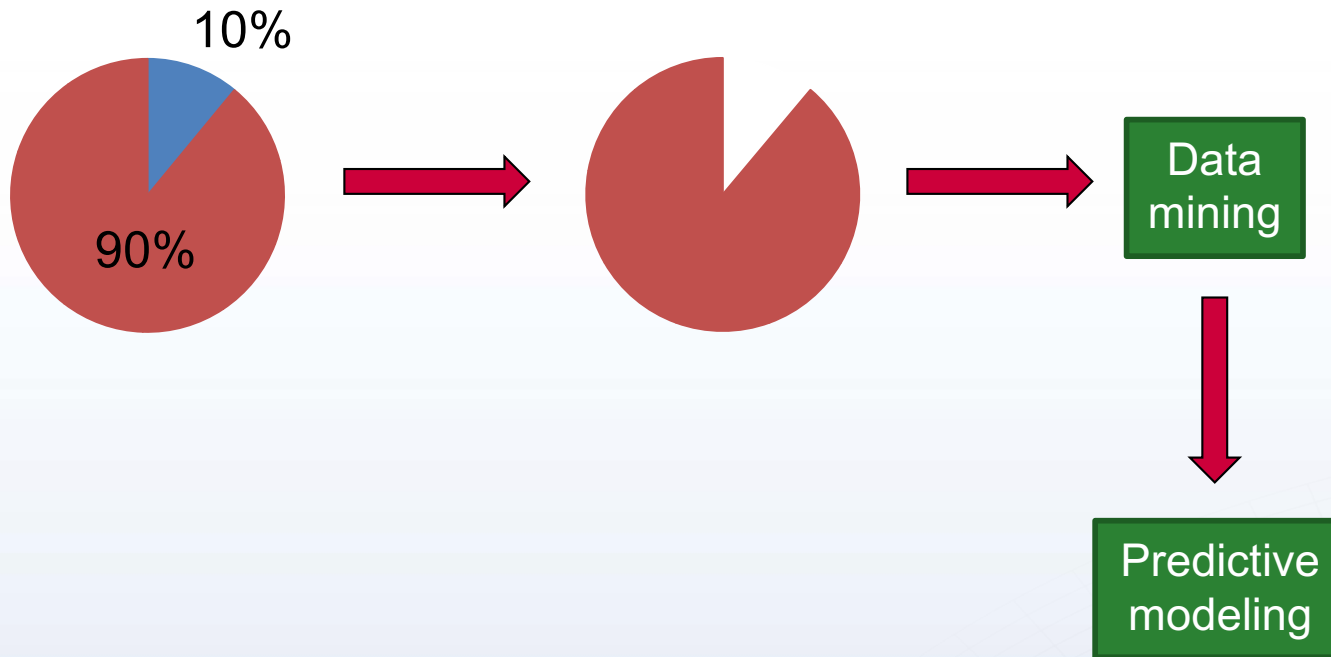
Cross-Validation



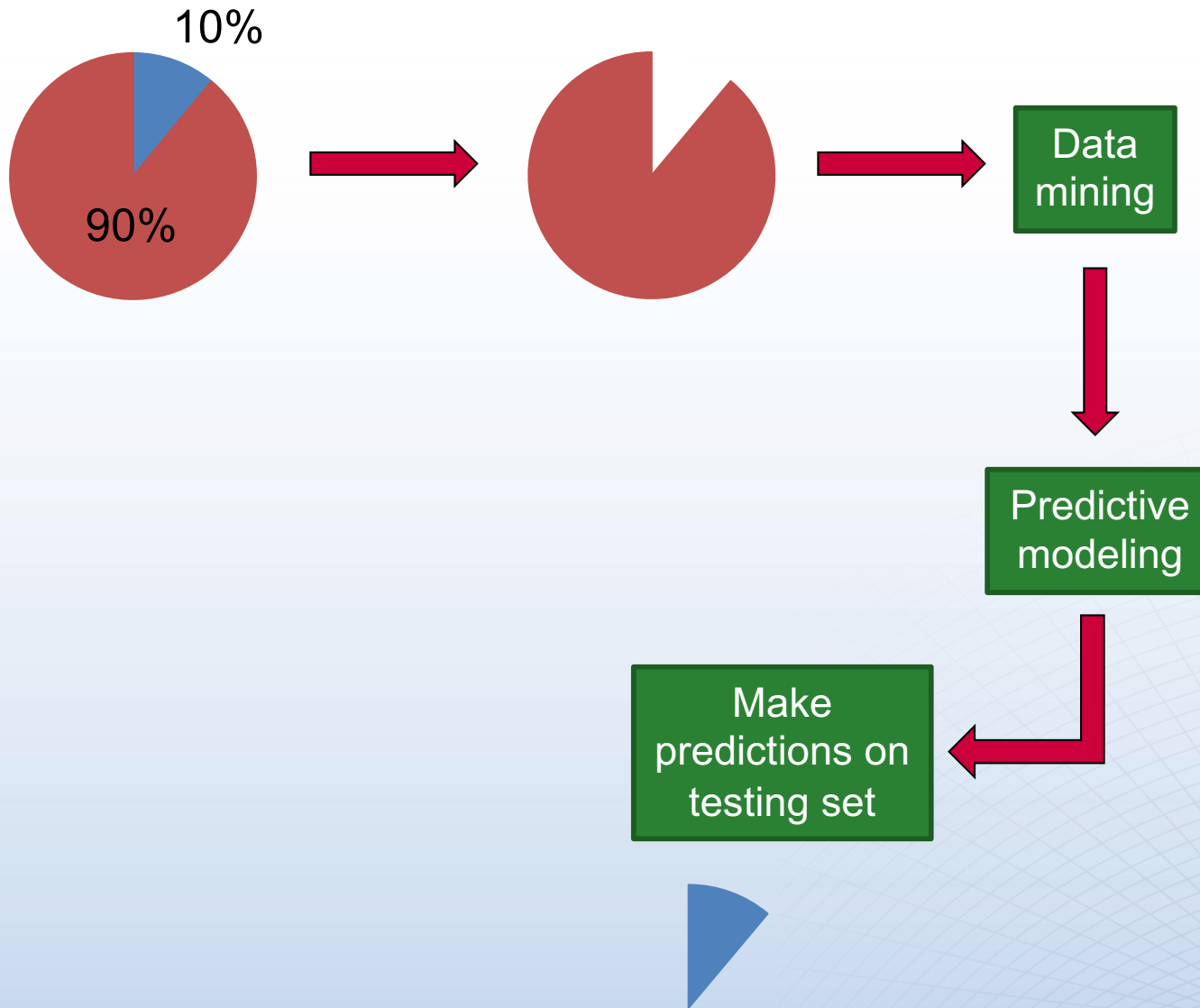
Cross-Validation



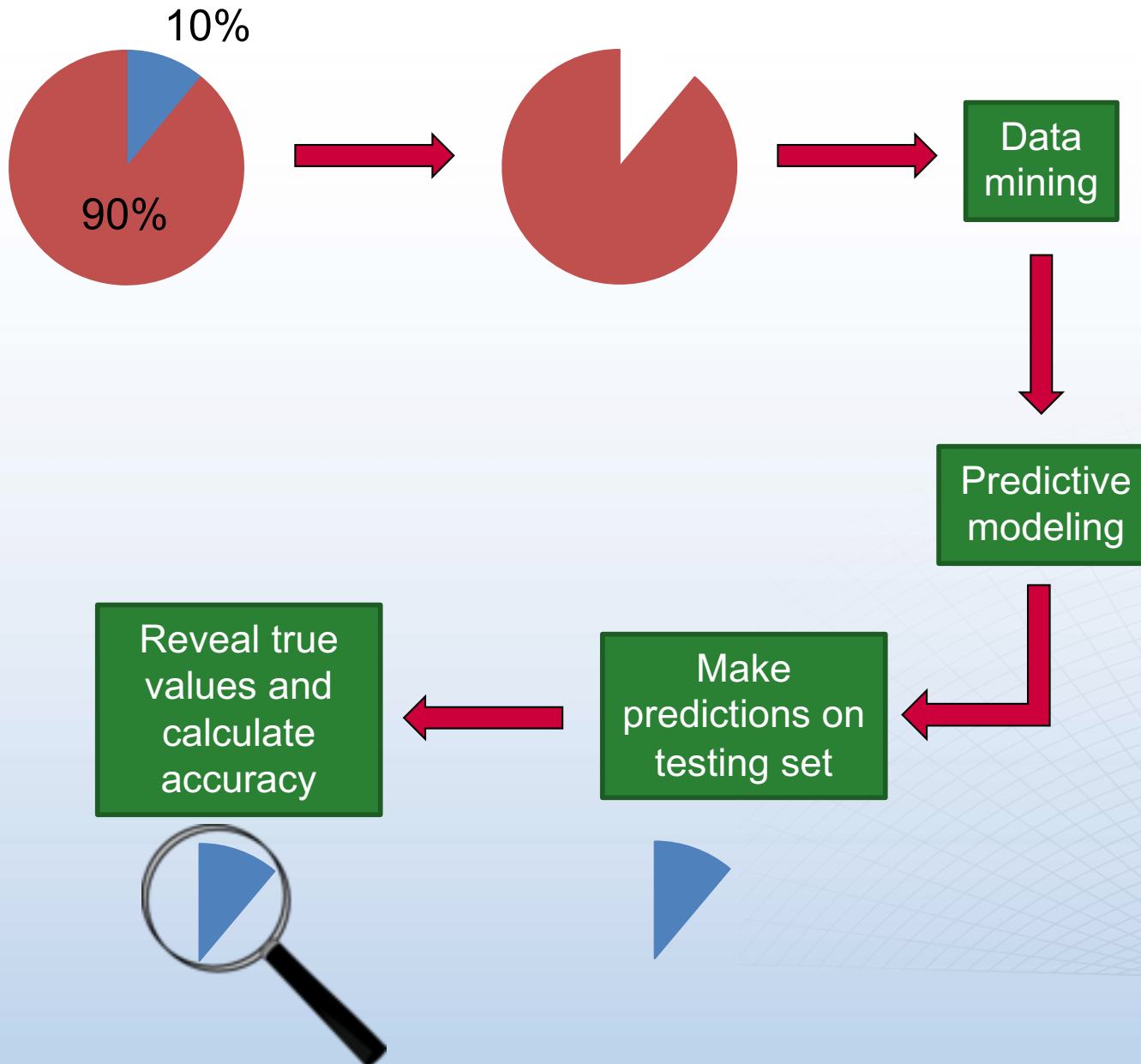
Cross-Validation



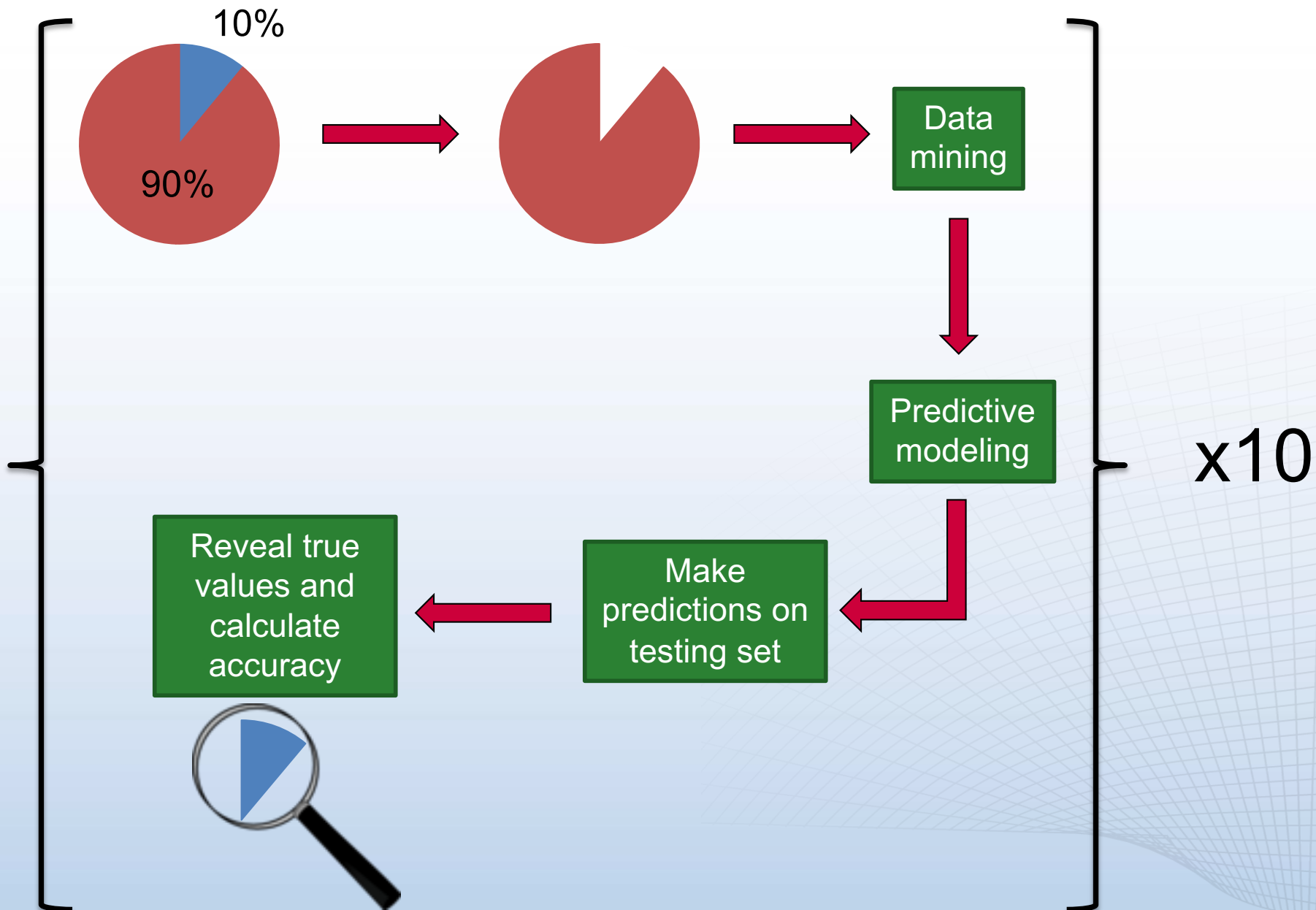
Cross-Validation



Cross-Validation



Cross-Validation

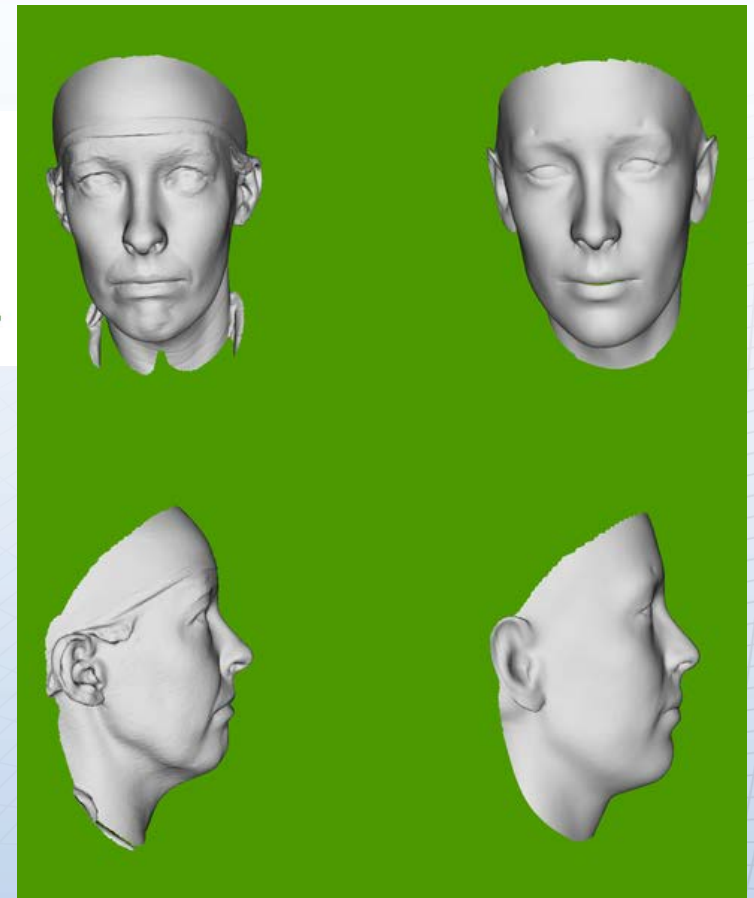
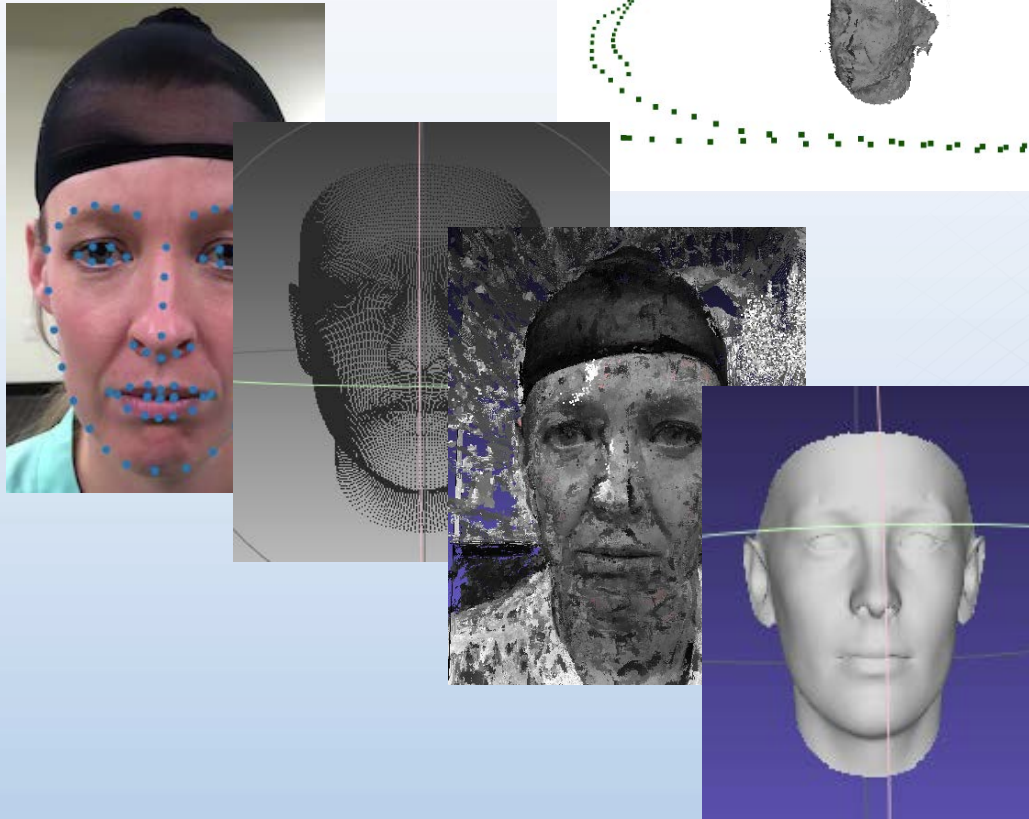
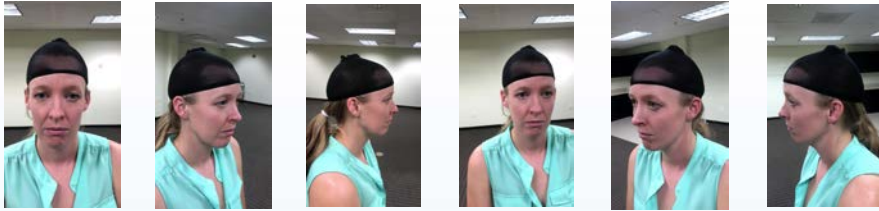


Cross-Validation

- › This means we are performing data mining and predictive modeling 10 times for each phenotype
- › However, we now have out-of-sample predictions on every single subject in our dataset
- › At the end, we build a final model using all of the data, and the cross-validation accuracy approximates the accuracy of this final model
- › Use the cross-validation results to make confidence statements about new predictions and *exclude* trait categories that are highly unlikely

Face Morphology

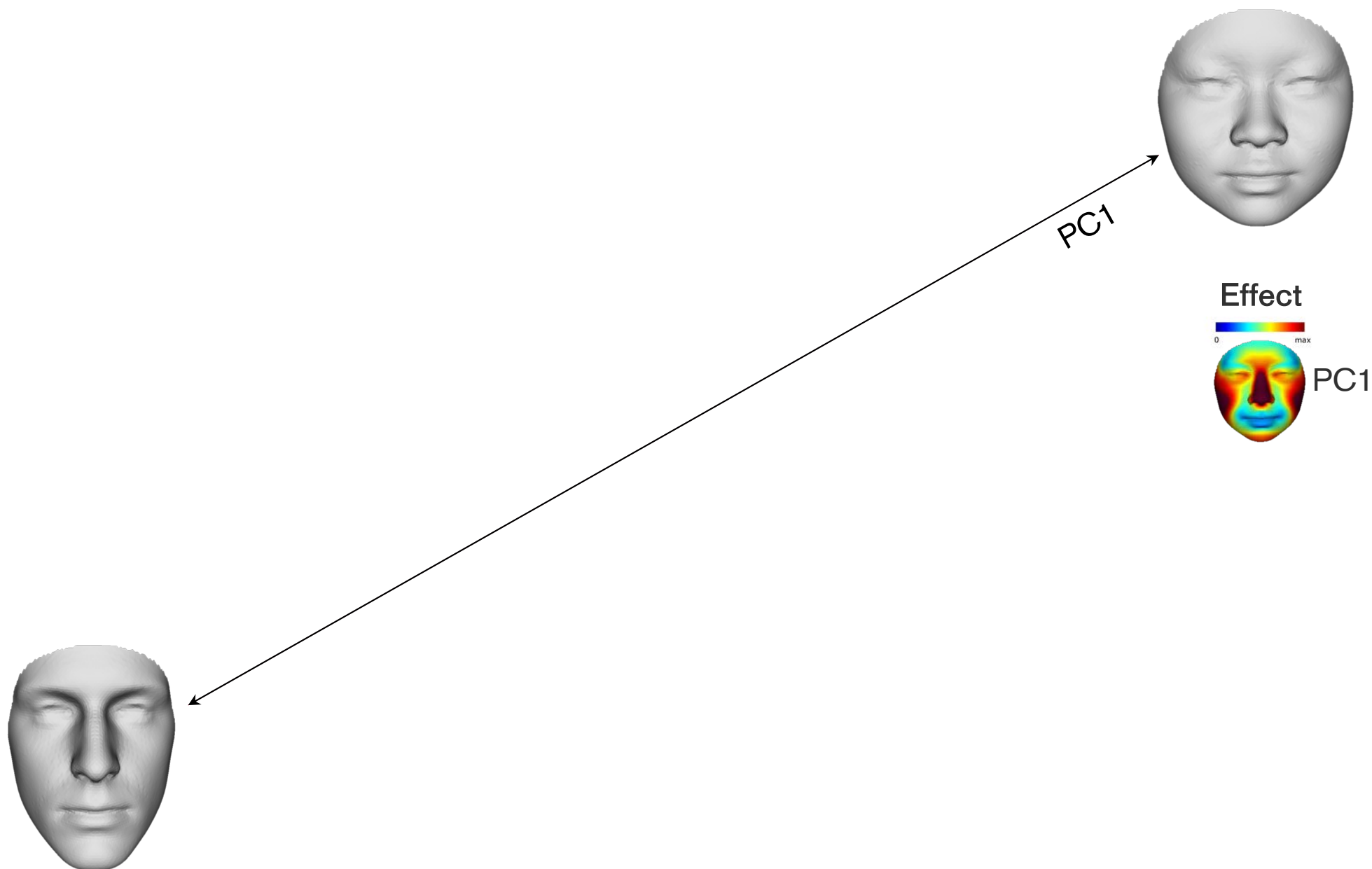
- > Can we apply this same process to a complex phenotype, such as face morphology?



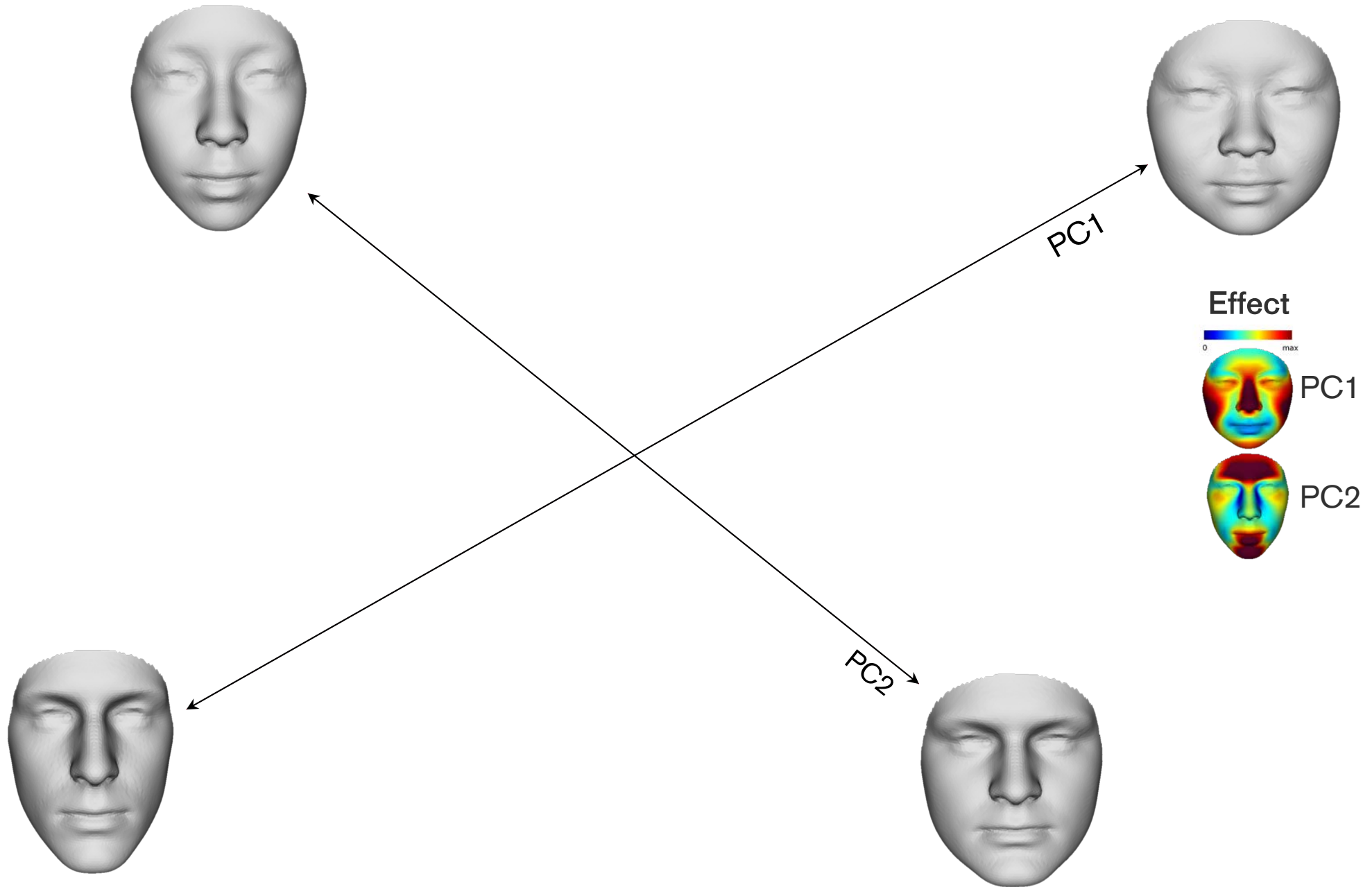
Face Morphology

- › Face morphology is captured as a 3D face scan using a smartphone app
- › The scan's shape is converted into thousands of quasi-landmarks (QLs), which are homologous across subjects
- › Each QL has 3 coordinates (x,y,z) , so faces are described by tens of thousands of numerical variables, many of which are correlated with one another
- › Principal component analysis (PCA) reduces the number of variables to a reasonable number while still retaining most of the variation

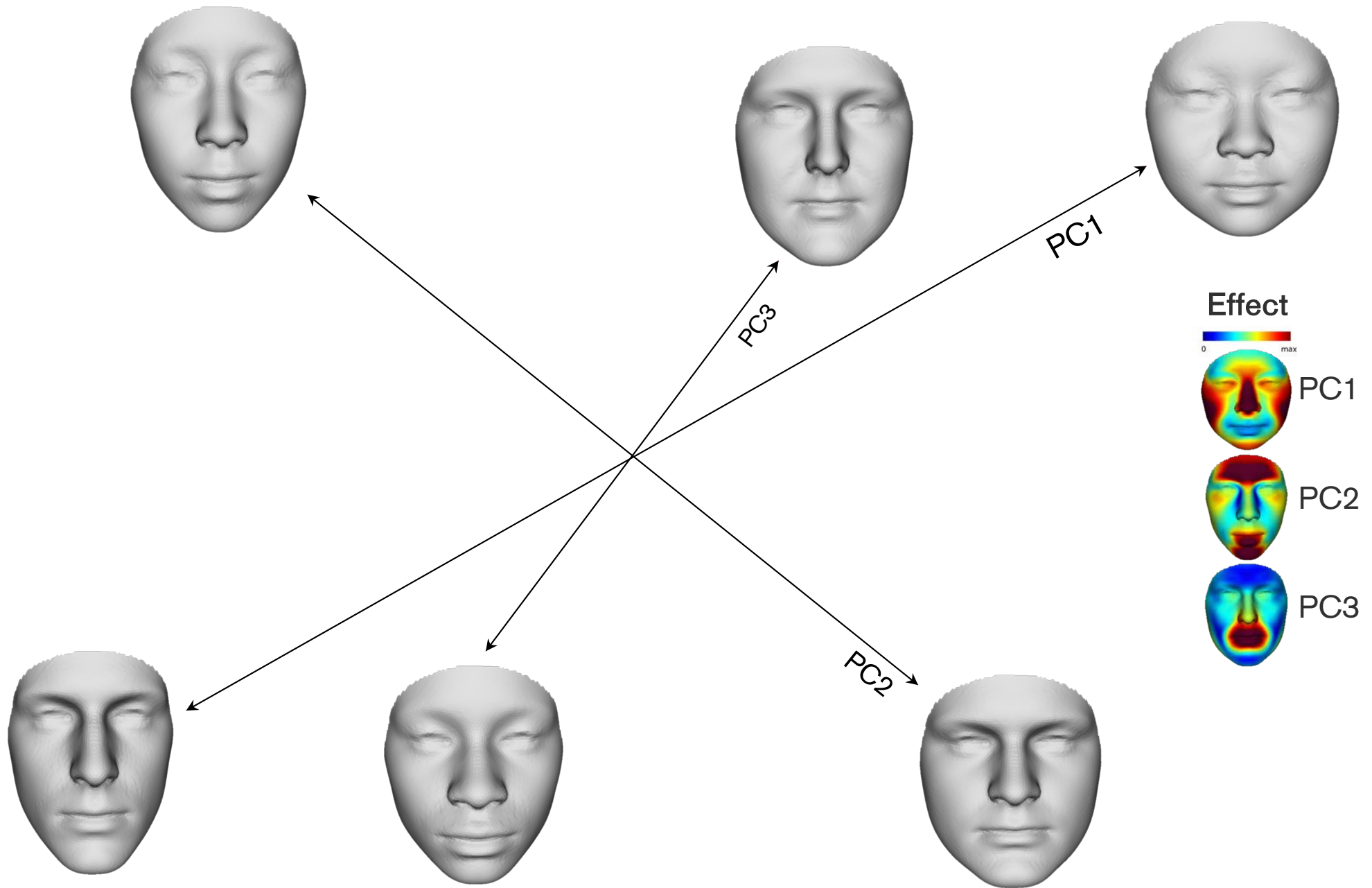
Face Space



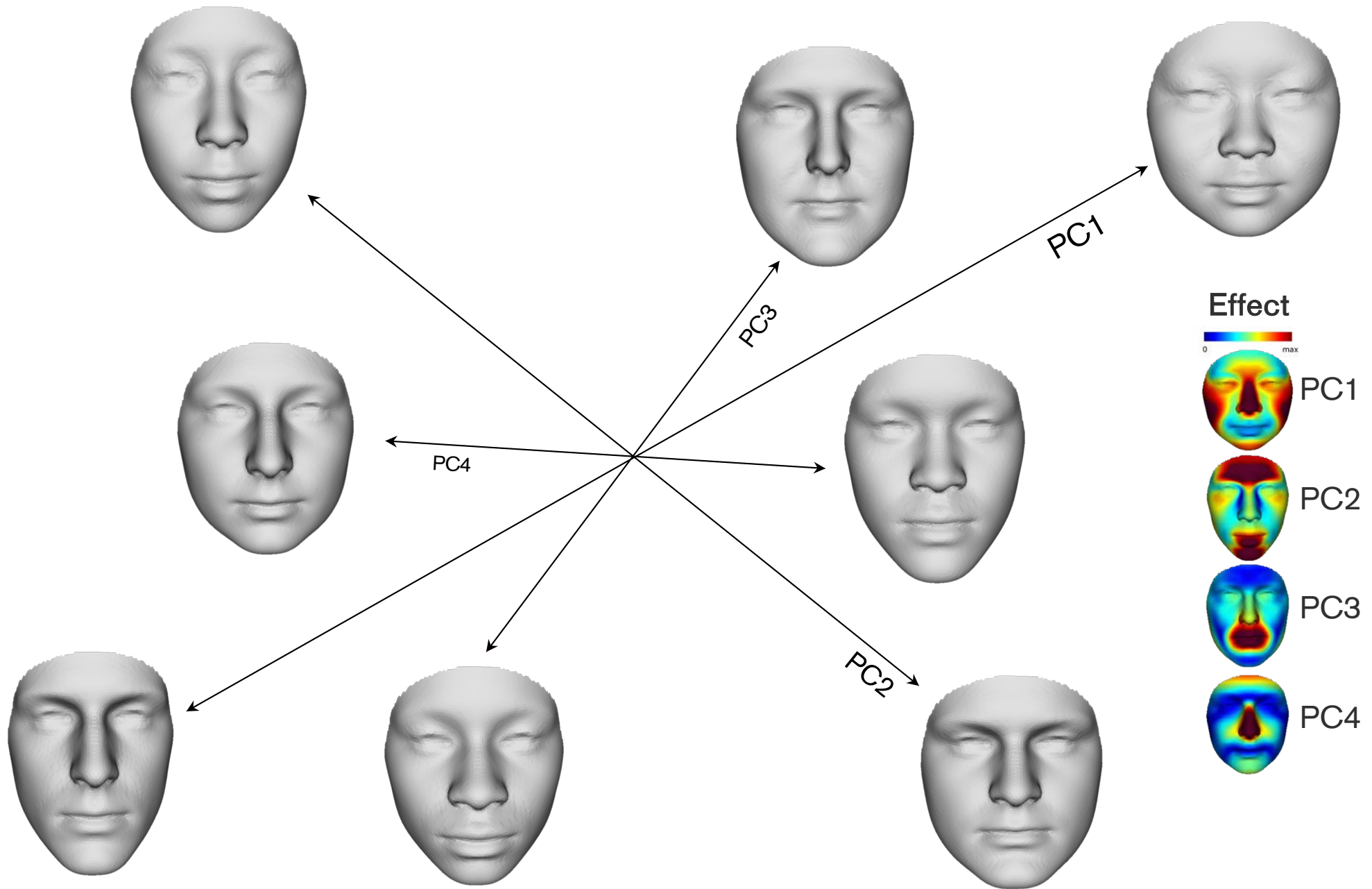
Face Space



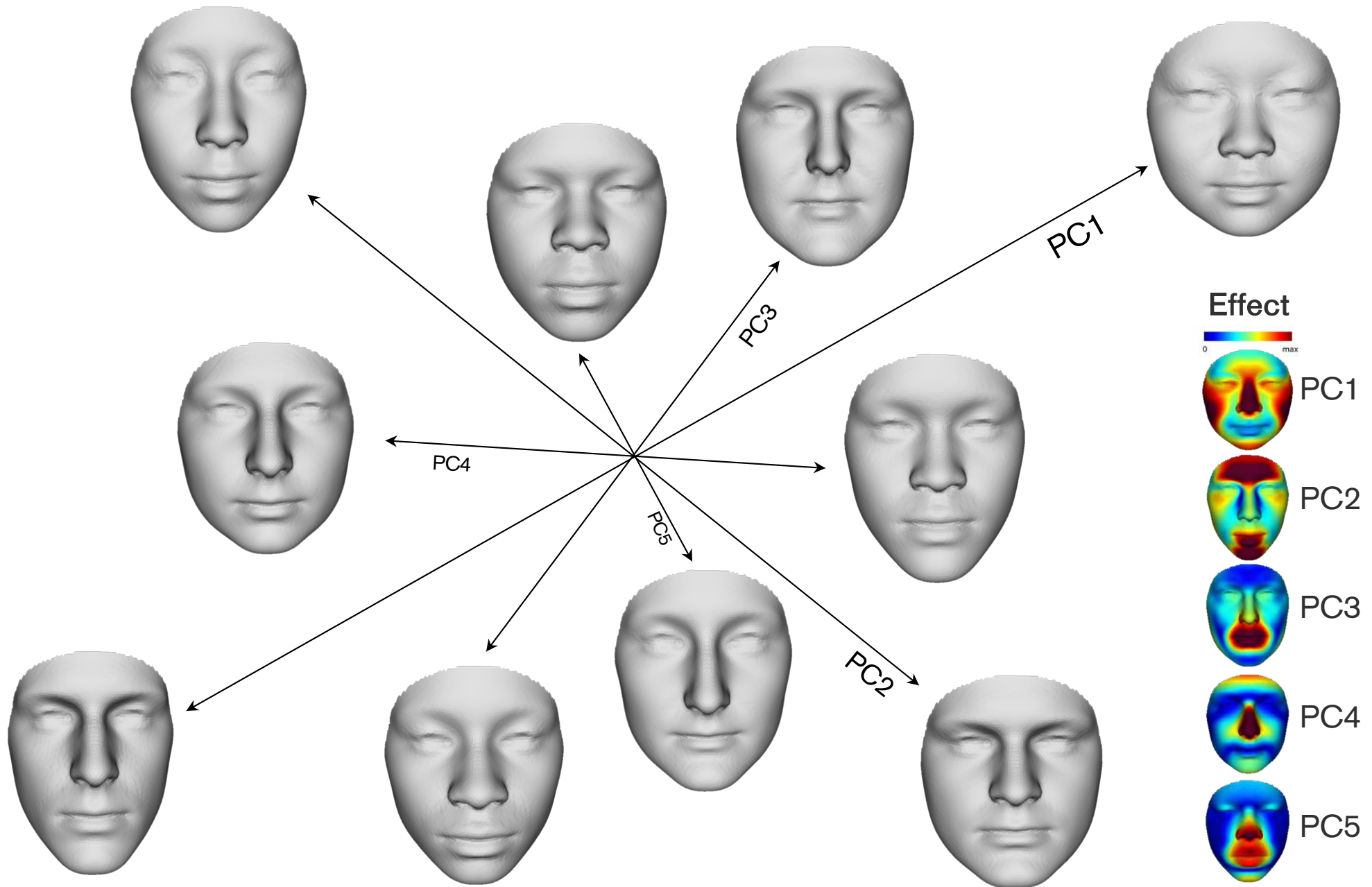
Face Space



Face Space



Face Space

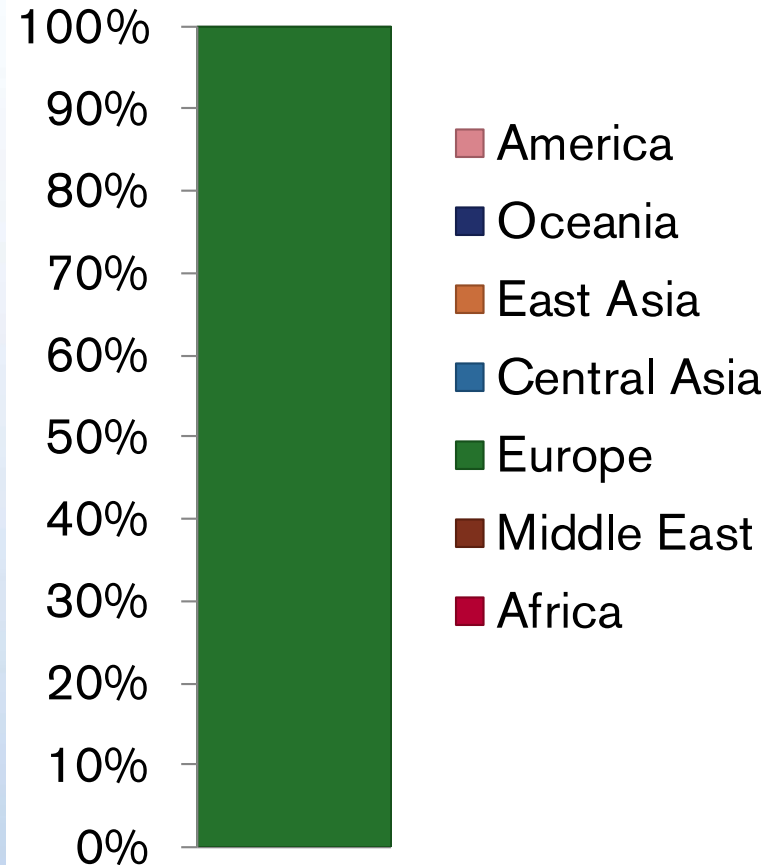


Snapshot: Making Predictions



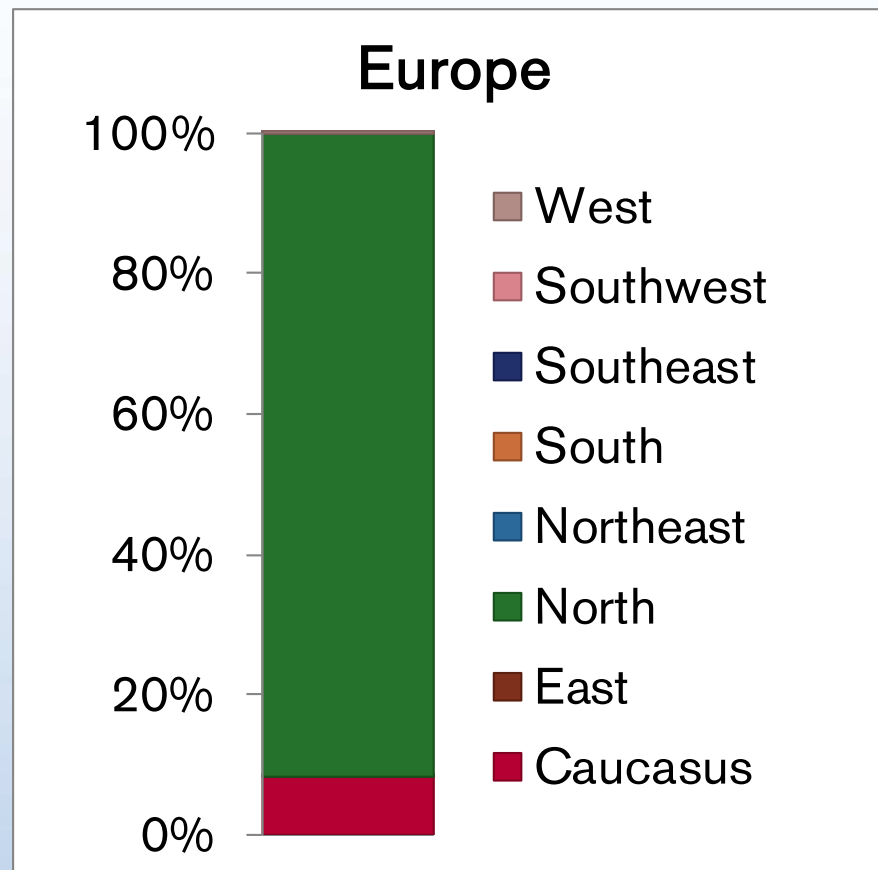
Global Ancestry

- › DNA from an unknown subject is partitioned into 7 parts according to its proportional similarity to each of the continental groups



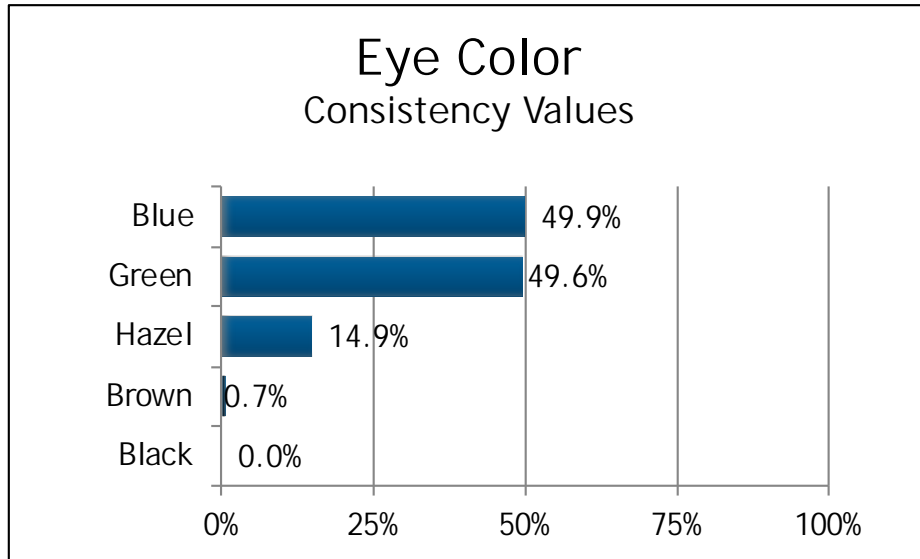
Regional Ancestry

- › Within each of the 7 continental populations, we can narrow the source of an individual's ancestry using the same statistical approach with different defined populations



G2P Prediction

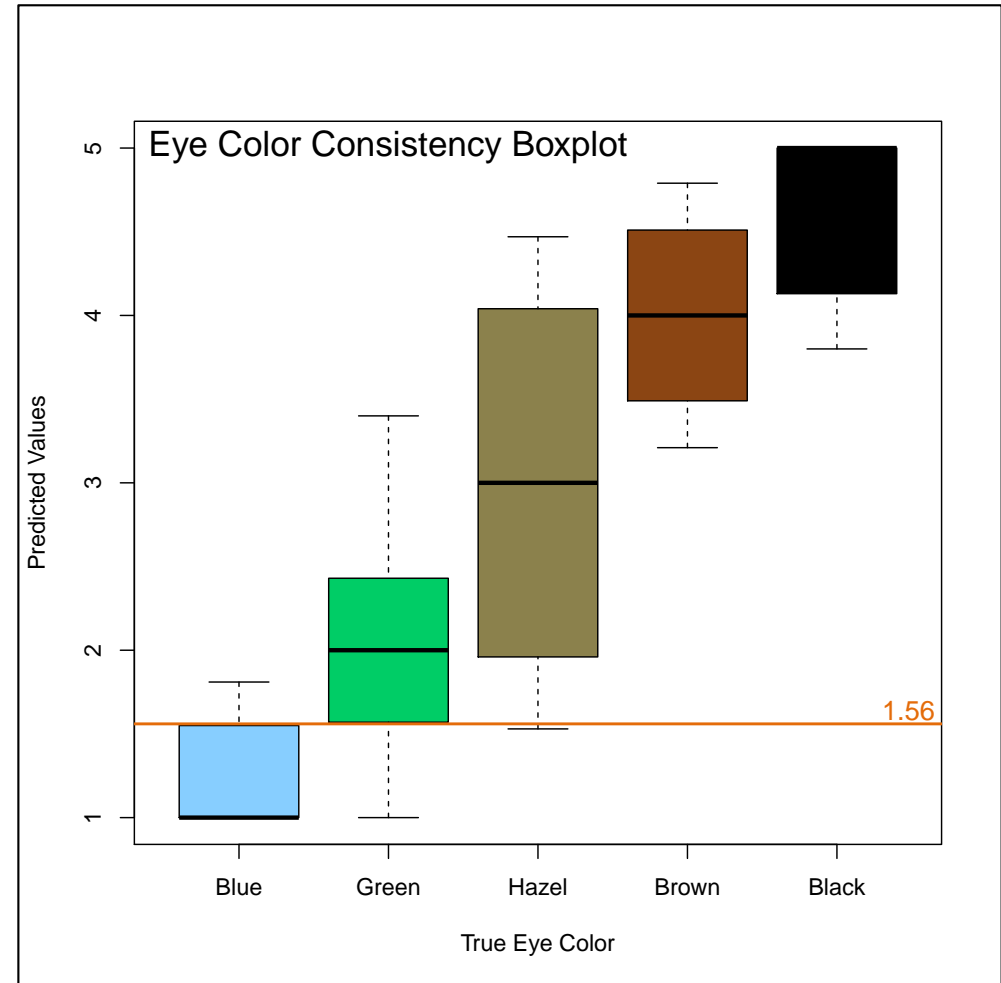
› Predicted Value = 1.560



› Blue (50.4% confidence)

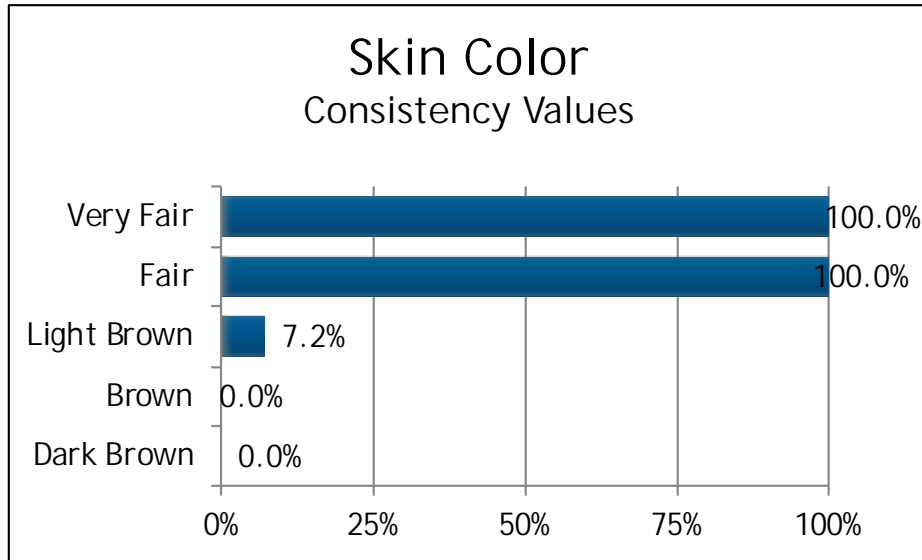
› Blue or Green (85.1% confidence)

› NOT Brown or Black (99.3% confidence)



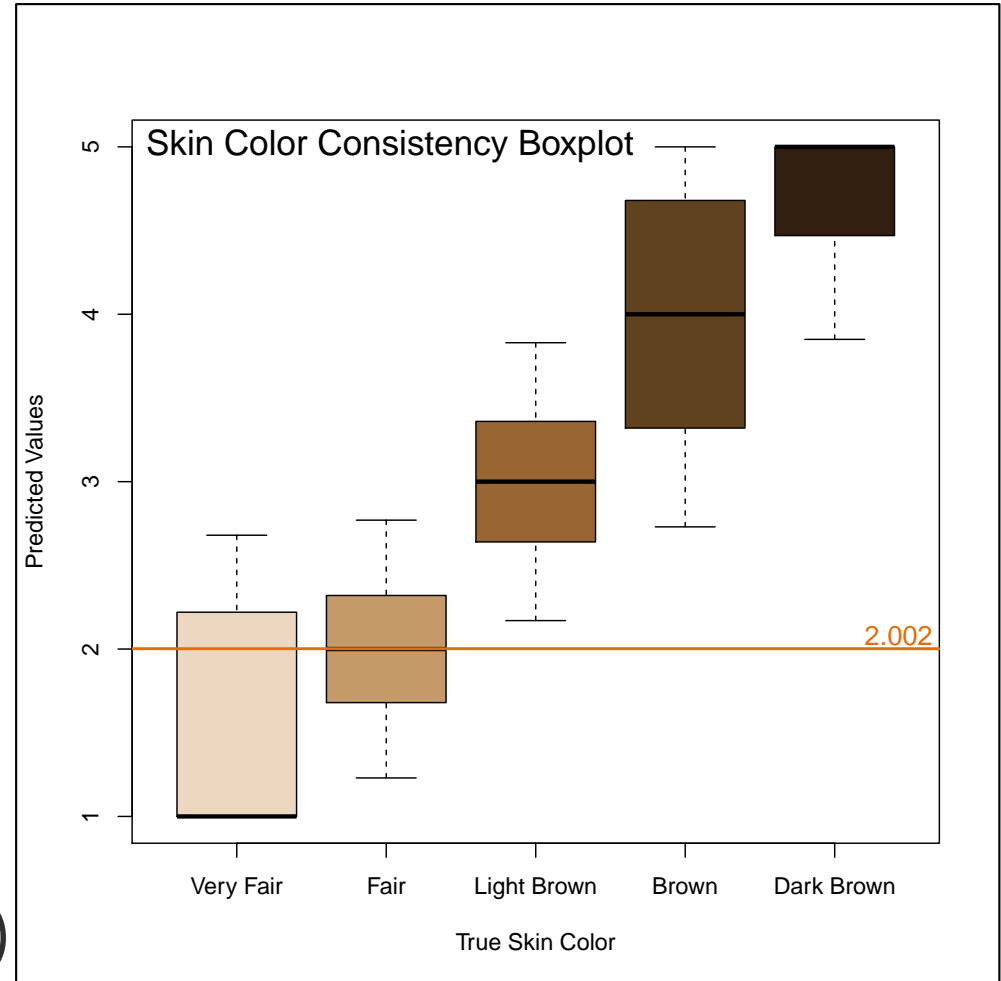
Sample Results – Skin Color

› Predicted Value = 2.002



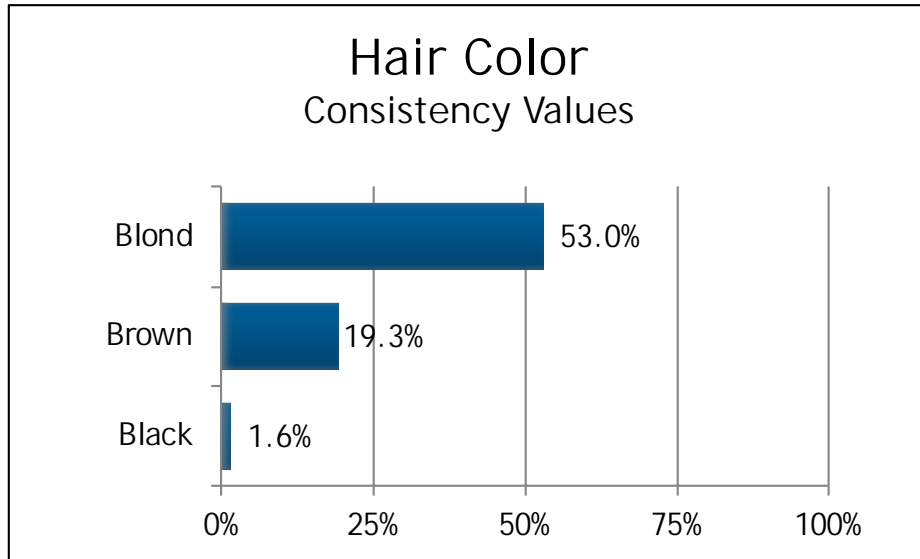
› Very Fair or Fair (92.9% confidence)

› NOT Light Brown, Brown, or Dark Brown (92.9% confidence)

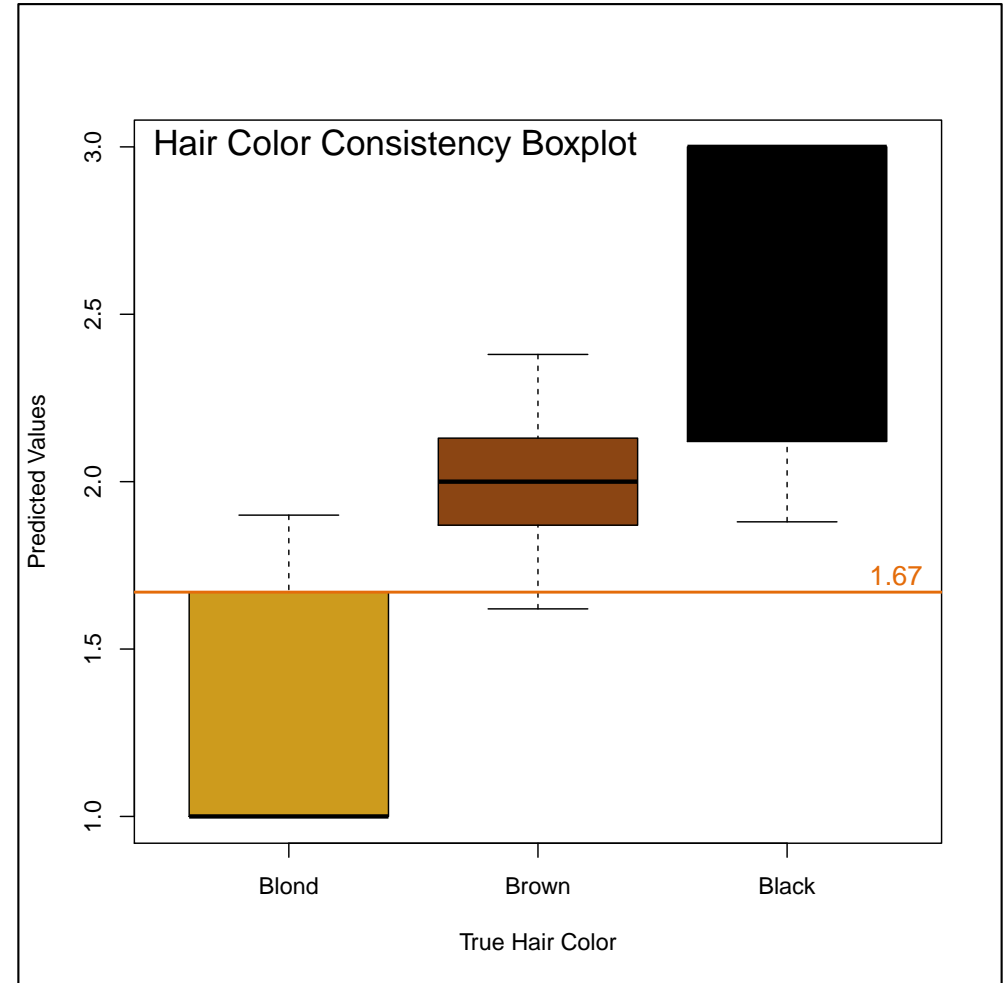


Sample Results – Hair Color

› Predicted Value = 1.670

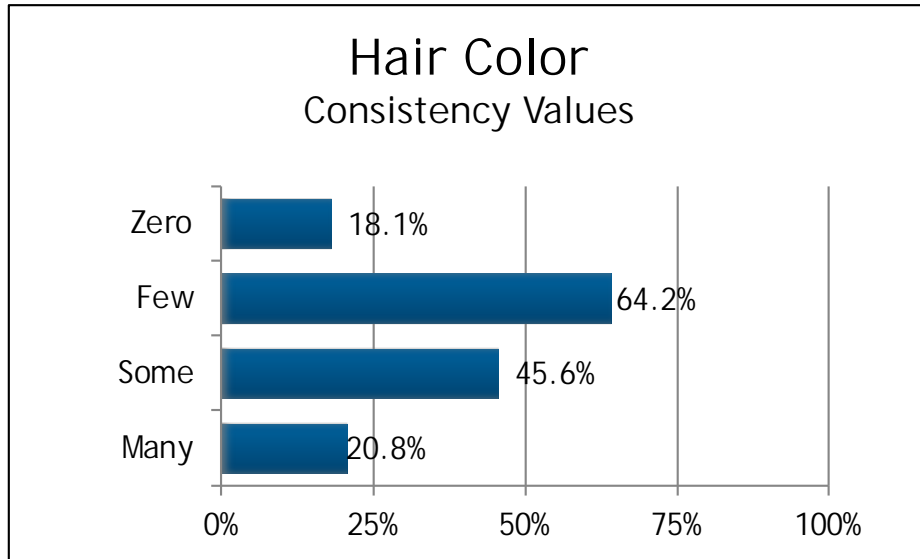


- › Blond (80.7% confidence)
- › Blond or Brown (98.3% confidence)
- › NOT Black (98.4% confidence)



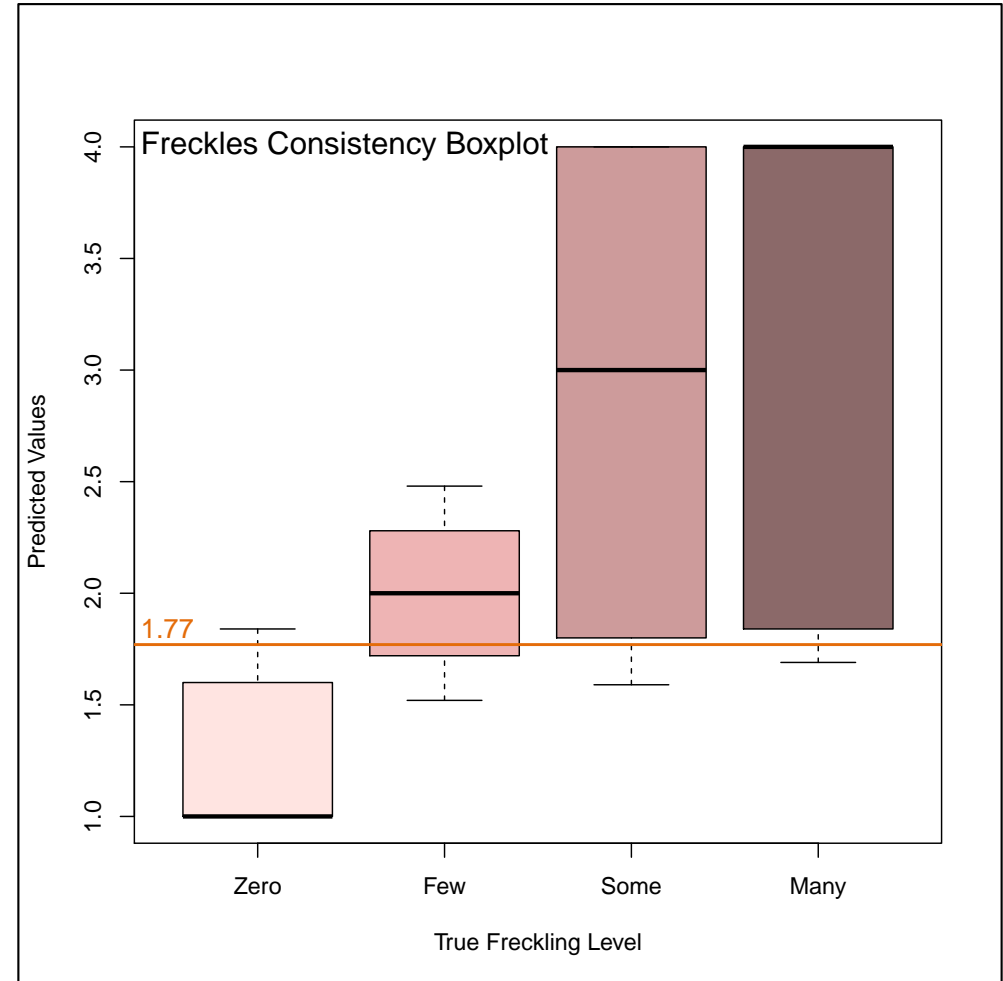
Sample Results – Freckling

› Predicted Value = 1.770



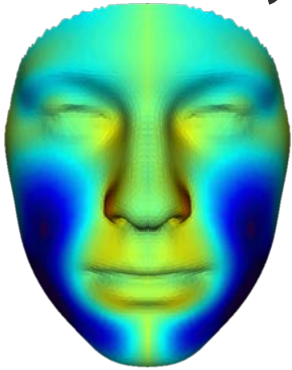
› Few (54.3% confidence)

› Few or Some (79.2% confidence)



Sample Results – Face Shape

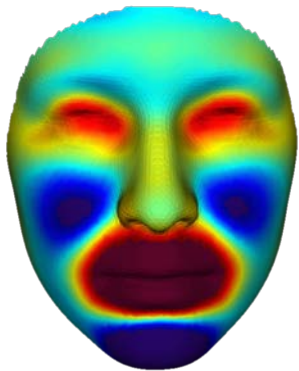
- Predict the face and compare it to a face predicted using only sex and ancestry; heat maps show the differences



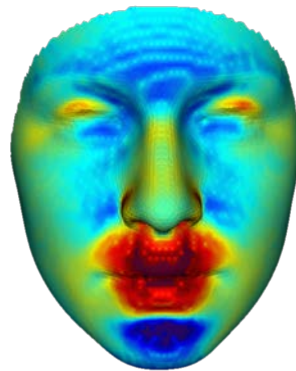
x: narrower jaw & chin;
slightly wider nose
& mouth



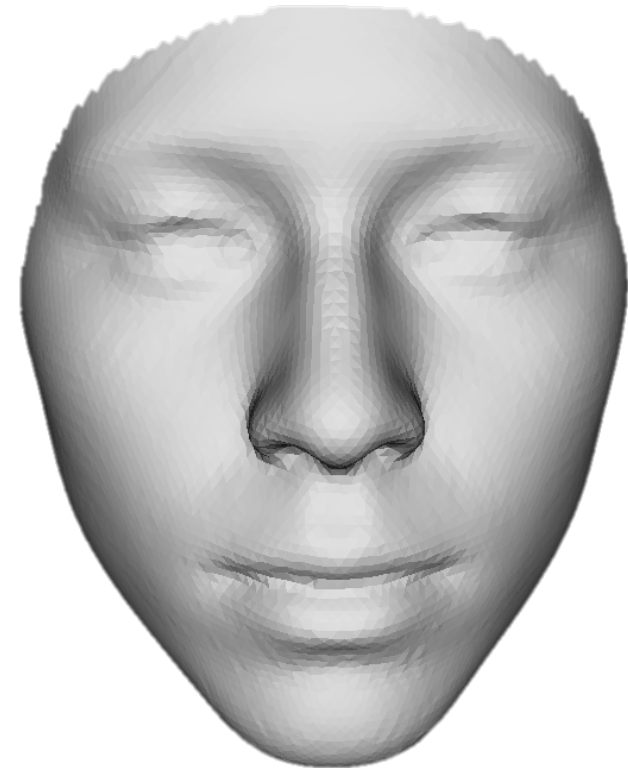
y: higher brow, nose
& jaw; lower mouth
& chin



z: more protruding
eyes & mouth; more
recessed cheeks & chin

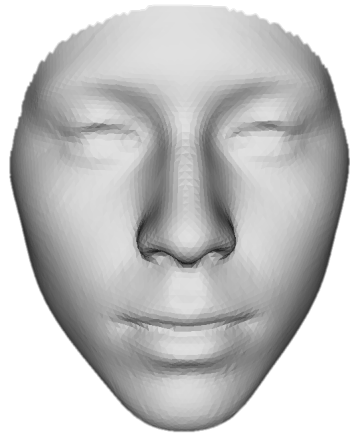


area: larger mouth;
smaller brow, cheeks
& chin



Sample Results – Face Shape

- Caricaturize the face according to the heat maps, fit a head to it, and apply the predicted pigmentation
 - Default age (25) and BMI (22)



Very Fair / Fair (92.8% confidence)
NOT Lt Brown / Brown / Dk Brown (92.8% confidence)



Blue / Green (85.1% confidence)
NOT Brown / Black (99.3% confidence)



Blond / Brown (98.4% confidence)
NOT Black (98.4% confidence)



Few / Some (79.2% confidence)

Sex: Female ♀

Age: Unknown
(Composite shown at age 25)

Body Mass: Unknown
(Composite shown at BMI 22, Normal)

Ancestry: Northern European



Law Enforcement Casework

- › Predictive models are built using high-quality data, so they assume that the data they receive when making a prediction is also high-quality
- › In casework, the DNA is of varying quality and quantity, and we need to take this into account when making predictions

Source		Type		Quantity		Call Rate	
Semen	48.3%	Single Source	80.2%	≤ 2.5 ng	22.0%	> 95%	47.9%
Blood	24.0%	Low Mixture	15.5%	2.5-5 ng	12.9%	90-95%	11.3%
Tissue	10.3%	High Mixture (Deconvoluted)	4.3%	5-10 ng	13.3%	80-90%	17.9%
Saliva	7.9%			10-20 ng	17.8%	70-80%	6.2%
Bone	5.0%			20-40 ng	27.0%	60-70%	12.1%
Touch	4.5%			40-80 ng	3.3%	<60%	4.7%
				>80 ng	3.7%		

Law Enforcement Casework

- › Genotyping call rates are never 100% with casework samples because of the small DNA quantities
 - We use a machine learning algorithm that allows for missing data (many do not)
 - For each sample, we recalculate the cross-validation results and confidence intervals for that set of SNPs
- › DNA degradation results in short fragment lengths
 - Optimized genotyping protocol for forensic samples
- › Mixtures confuse both genotype calling and phenotype prediction
 - We developed new computational methods to deconvolute mixtures

Law Enforcement Casework

Snapshot Prediction Results
Composite Profile

IC90-FI-2001-180951-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 98.0
Dark / Dk Olive (99.99% confidence)
Light (0.01% confidence)

Sex: Male ♂
Age: Unknown

Snapshot Prediction Results
Composite Profile

IC920418-Agawam-MA-Snapshot-R01

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 4.8
Fair / Very Fair (90.4% confidence)
Light (9.6% confidence)

Sex: Male ♂
Age: Unknown

Snapshot Prediction Results
Composite Profile

Snapshot #RM1305293

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 17.3
Fair / Very Fair (84.4% confidence)
Light (15.6% confidence)

Sex: Male ♂
Age: Unknown

Snapshot Prediction Results
Composite Profile

Snapshot #MA015962-889013861

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 2.7
Very Fair / Fair (98.2% confidence)
Light (1.8% confidence)

Sex: Male ♂
Age: Unknown

Snapshot Prediction Results
Composite Profile

4K958-8-2016-02105-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 5.1
Fair / Very Fair (89.3% confidence)
Light (10.7% confidence)

Sex: Male ♂
Age: Unknown
Height progressed to -40 years

Eye Color: 19.5
Green / Hazel (99.3% confidence)
Light (0.7% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Snapshot Prediction Results
Composite Profile

4LPD-IL-76-00714-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 18.0
Light Brown / Fair (77.4% confidence)
Light (22.6% confidence)

Sex: Male ♂
Age: Unknown
(Composite shown at age 29)

Eye Color: 22.3
Green / Hazel (99.0% confidence)
Light (1.0% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Snapshot Prediction Results
Composite Profile

30 WA-INCMEC-01015-IND-1161321018-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 5.8
Fair / Very Fair (88.0% confidence)
Light (12.0% confidence)

Sex: Female ♀
Age: Unknown
(Composite shown at age 53)

Eye Color: 72.4
Hazel / Brown (99.5% confidence)
Light (0.5% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Snapshot Prediction Results
Composite Profile

4LAS058-72-2077-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 19.8
Fair / Very Fair (82.1% confidence)
Light (17.9% confidence)

Sex: Male ♂
Age: Unknown
(Age progressed to -69 years)

Eye Color: 85.0
Brown / Hazel (99.2% confidence)
Light (0.8% confidence)

Body Mass: Unknown
(Composite shown at BMI 22, Normal)

Snapshot Prediction Results
Composite Profile

4S30-FI-99-29040-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 4.7
Fair / Very Fair (97.8% confidence)
Light (2.2% confidence)

Sex: Male ♂
Age: Unknown
(Composite shown at age 26)

Eye Color: 71.3
Hazel / Brown (88.8% confidence)
Light (11.2% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Hair Color: 19.3
Black / Brown (99.8% confidence)
Light (0.2% confidence)

Ancestry: Admixed European (West/South) & North Middle Eastern

Freckles: 63.4
Few / Some (79.3% confidence)
None (20.7% confidence)

Snapshot Prediction Results
Composite Profile

4THPD-96-12-1073-Snapshot-R01

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 5.5
Fair / Very Fair (96.8% confidence)
Light (3.2% confidence)

Sex: Male ♂
Age: Unknown
(Composite shown at age 25)

Eye Color: 19.3
Green / Hazel (75.8% confidence)
Light (24.2% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Ancestry: Northern European

Hair Color: 14.0
Blond / Brown (98.4% confidence)
Light (1.6% confidence)

Freckles: 96.3
Few / Many (91.8% confidence)
None (8.2% confidence)

Snapshot Prediction Results
Composite Profile

4BBDP-FI-98-02885-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 6.0
Fair / Very Fair (93.6% confidence)
Light (6.4% confidence)

Sex: Male ♂
Age: Unknown
(Age progressed to -40 years)

Eye Color: 10.3
Green / Hazel (74.1% confidence)
Light (25.9% confidence)

Body Mass: Unknown
(Composite shown at BMI 22)

Ancestry: Northern, Central and/or Southeast European

Hair Color: 11.4
Blond (93.7% confidence)
Light (6.3% confidence)

Freckles: 31.9
Zero / Few (99.5% confidence)
Some (0.5% confidence)

Snapshot Prediction Results
Composite Profile

4YCO-MI-73-16429-Snapshot

Predicted (☑) & Excluded (☒) Phenotypes

Skin Color: 13.3
Fair / Very Fair (90.4% confidence)
Light (9.6% confidence)

Sex: Male ♂
Age: Unknown
(Composite shown at age 23)

Eye Color: 17.3
Green / Hazel (75.3% confidence)
Light (24.7% confidence)

Body Mass: Unknown
(Composite shown at BMI 22, Normal)

Ancestry: Northern and/or Central European

Hair Color: 23.1
Brown / Blond (97.9% confidence)
Light (2.1% confidence)

Freckles: 90.6
Some / Many (10.8% confidence)
None (89.2% confidence)

Law Enforcement Casework

Snapshot Prediction Results Composite Profile



#C920418-Agawam-MA-Snapshot-R01
PNL Document #17118Q38-9FTVRN6R



Actual Photo

Predicted (■) & Excluded (☒) Phenotypes



Sex: Male ♂

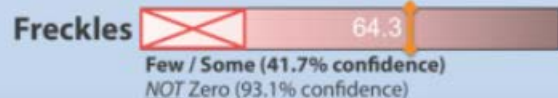
Age: Unknown
(Age progressed to ~50 years)



Body Mass: Unknown
(Composite shown at BMI 24)



Ancestry: European -
North/South Admixed,
West, or South



- › 1992 homicide of Lisa Ziegert in Agawam, MA
- › Thousands of men listed in the case file
- › Detectives used phenotyping results to prioritize who to talk to, starting with the few who most closely matched the predictions
- › One of those men fled, eventually confessing to the crime
- › His DNA matched the crime scene sample

Law Enforcement Casework

Snapshot Prediction Results Composite Profile

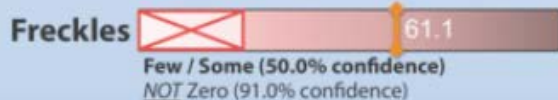
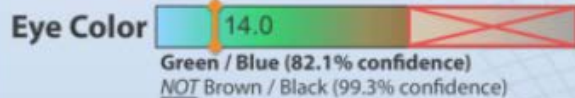


#SCL-09-4465-LA-Snapshot
PNL Document #17G25R40-GDC6R9U4



Driver's License

Predicted (■) & Excluded (☒) Phenotypes



Sex: Male ♂

Age: Unknown
(Composite shown at age 25)

Body Mass: Unknown
(Composite shown at BMI 22, Normal)

Ancestry: Northern European



- › 2009 homicide of Sierra Bouzigard in Lake Charles, LA
- › For 6 years, detectives had been looking for a Hispanic male based on Sierra's cell phone history
- › DNA phenotyping results were completely different
- › Detectives released the composite to the public
- › A tip led to a suspect whose DNA matched the crime scene sample

Distant Kinship Inference



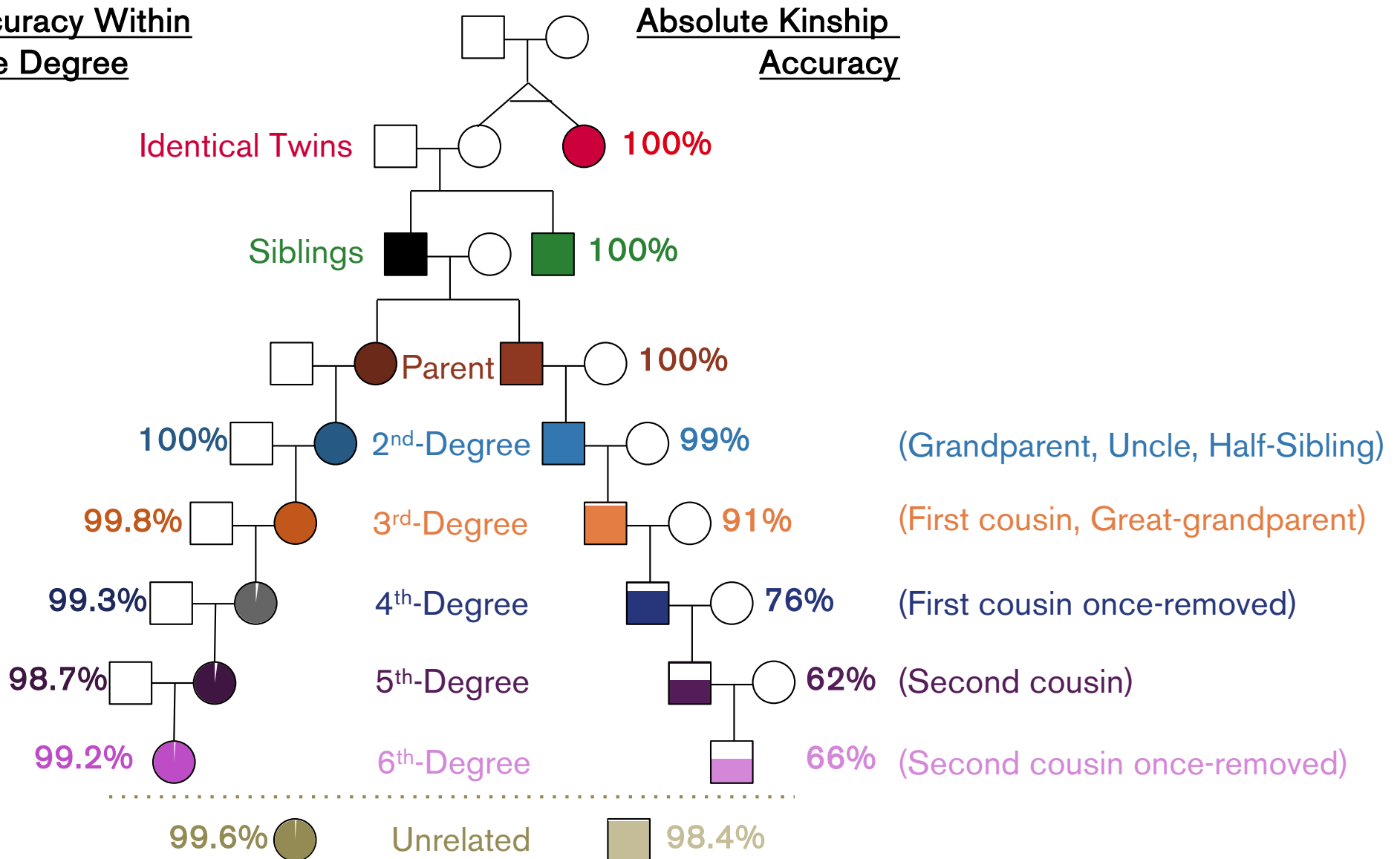
Kinship Inference

- › **DoD problem:** phenotype information isn't always enough
 - Enemy groups sometimes have a familial connection
 - If a DNA sample doesn't directly match known individuals in a database, might it instead be related to someone in that database?
- › **SBIR Modification:** Given DNA from two subjects, determine their degree of relatedness
- › Snapshot predictive models
 - Outcome is relationship between a pair of genomes
 - Supervised machine learning using thousands of related and unrelated pairs, including families with members of different ethnic backgrounds

Kinship Inference Accuracy

Accuracy Within One Degree

Absolute Kinship Accuracy



Kinship from Highly Degraded DNA

- › Snapshot Kinship requires high-quality data for tens of thousands of SNPs
- › Past accounting mission: identify missing soldiers from past conflicts by matching their DNA to family references
 - Autosomal and Y STRs: limited success on highly degraded samples, require family references with specific relationships
 - Mitochondrial DNA (mtDNA): requires family references with specific relationships (direct maternal relatives)
- › **DoD Problem:** Not all missing soldiers can be matched to their available family references using existing techniques
- › **Collaboration with AFMES-AFDIL:** autosomal kinship from highly degraded DNA
 - DoD Office of the Deputy Assistant Secretary of Defense for Emerging Capabilities and Prototyping

Kinship from Highly Degraded DNA

> Innovation:

- Laboratory: new methods to obtain thousands of autosomal SNPs from highly degraded bone samples
- Bioinformatics: determine kinship from sparse, low-quality data

Family Reference	A	A	T	G	C	A	T	G	G	T	A	C
	A	G	T	A	C	A	C	G	A	T	G	C
Unidentified	A	?	?	G	?	?	?	G	?	T	?	?
	A	?	C	?	?	?	?	?	A	C	?	?

- > Blind test: successfully matched samples from as far back as World War II to family references
 - Transitioning to AFDIL

Genetic Genealogy

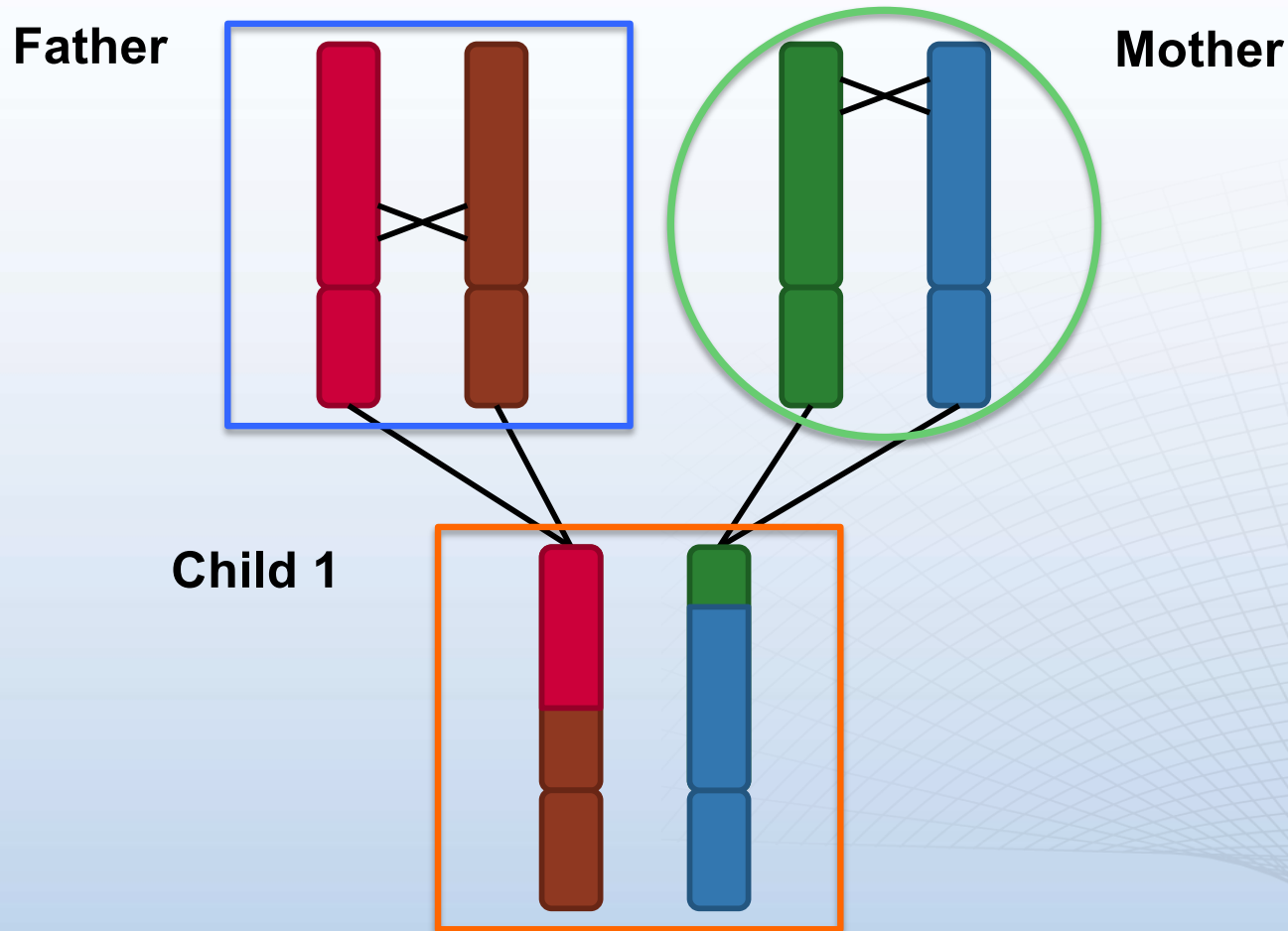


Procedure

1. Generate DNA data from a forensic sample from an unknown subject
2. Upload DNA data to a database to find other individuals that share DNA with the subject

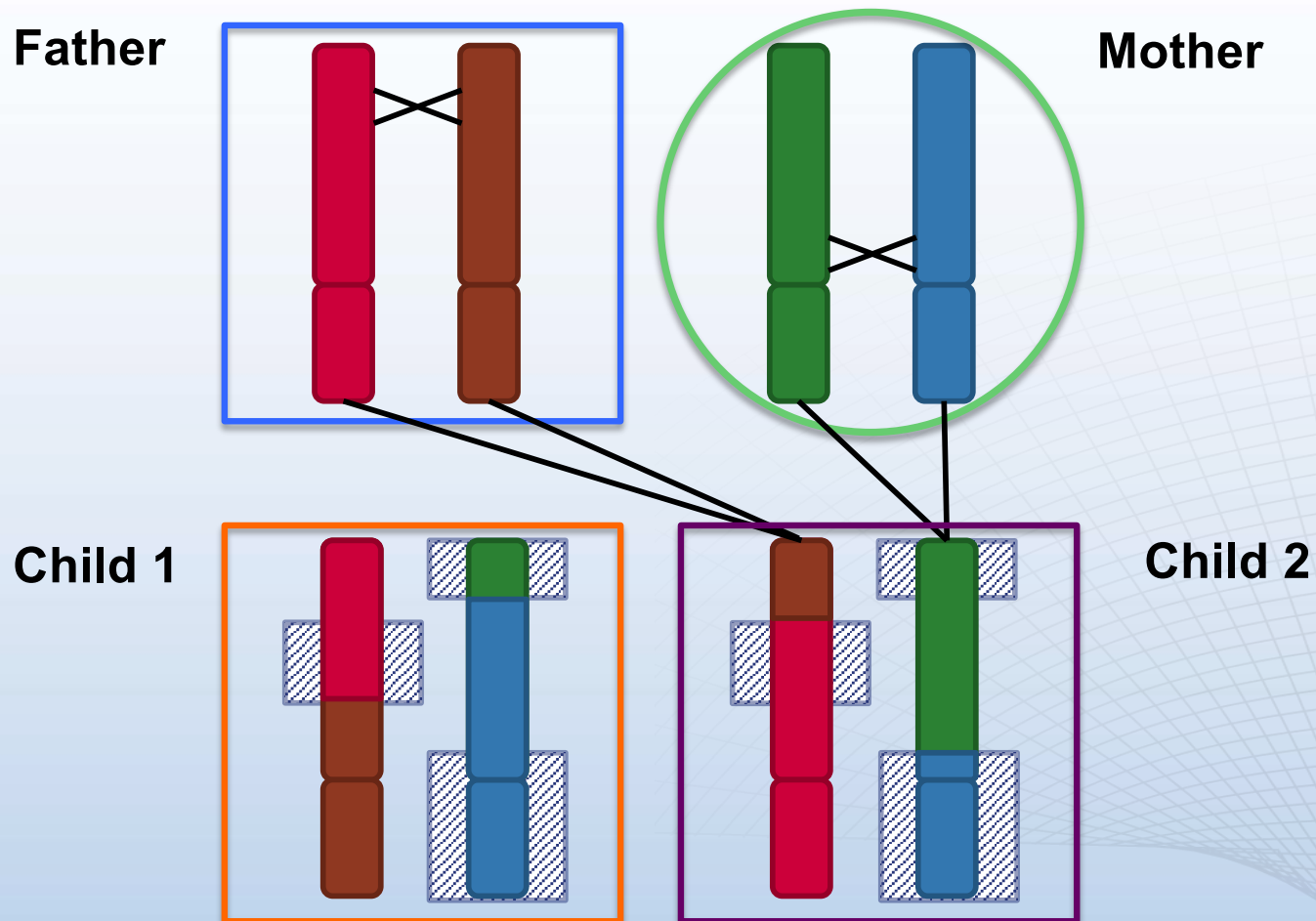
How does it work?

- › A child inherits one copy of each chromosome from their father and one from their mother
- › The chromosomes recombine, so each parent passes on a version that is a *combination* of their own two copies



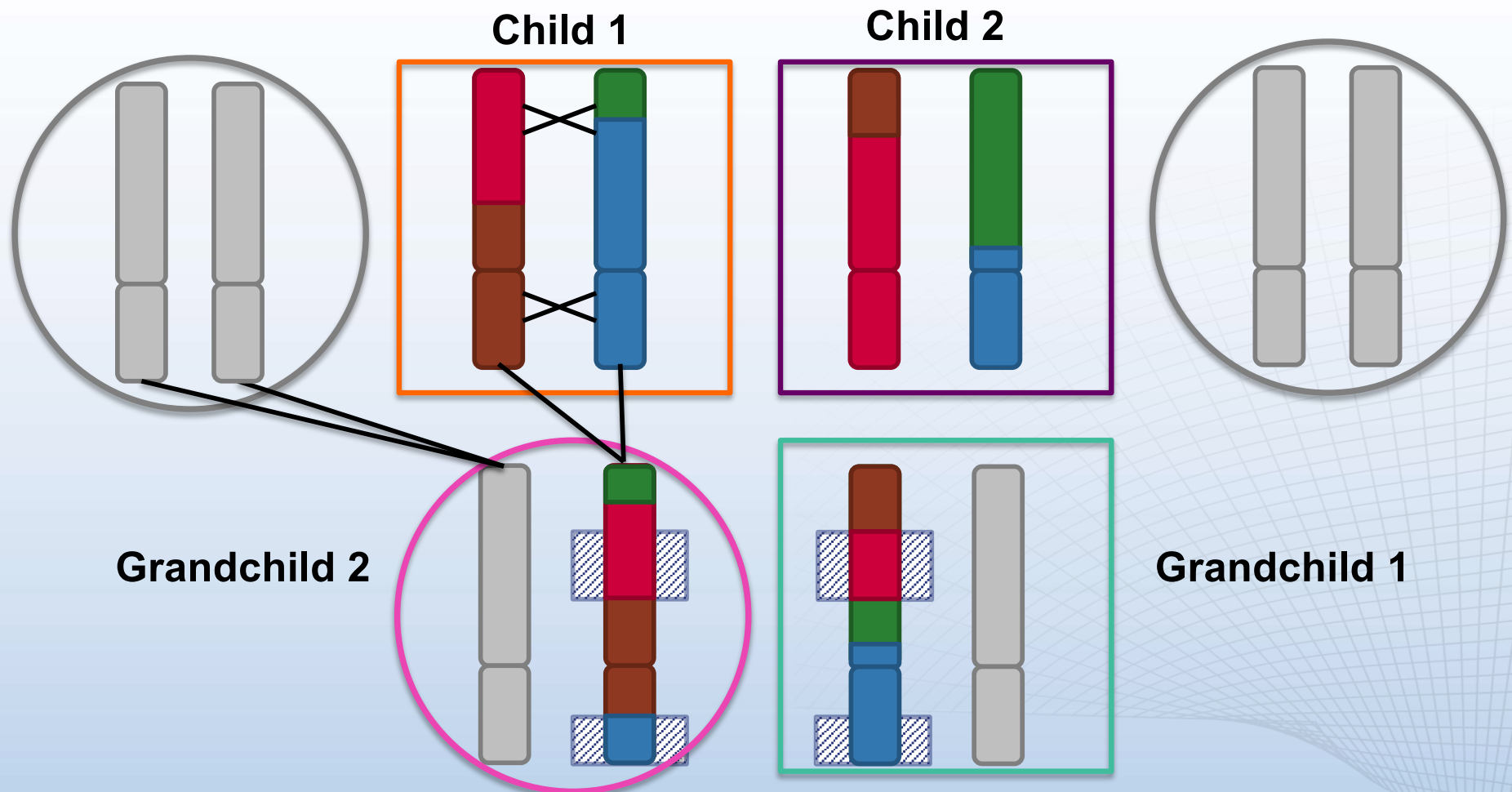
How does it work?

- › Recombination happens randomly for each child, so full siblings do not inherit the same copies



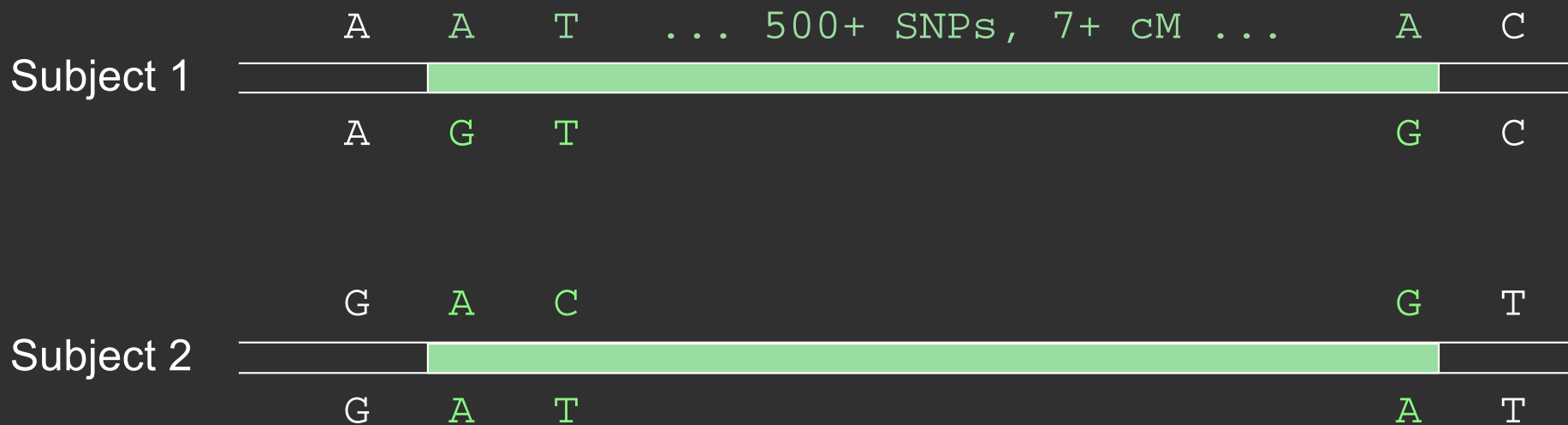
How does it work?

- > The more distantly related two subjects are, the smaller the shared segments become as they are broken up by recombination



Detecting Shared Segments

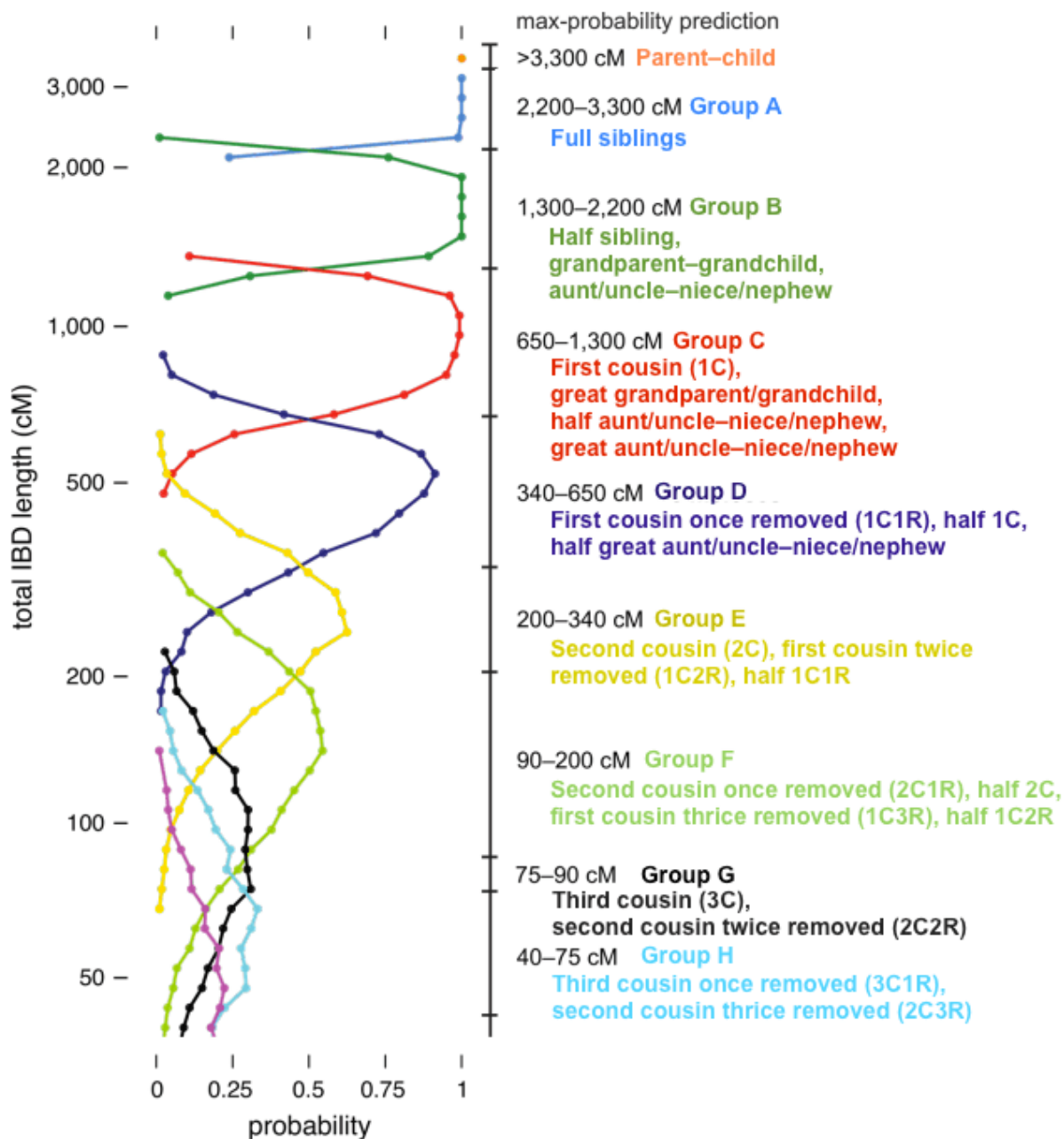
- › Common to share at least one allele at a SNP, but not common to have long uninterrupted stretches of shared alleles unless the two individuals inherited that DNA from a common ancestor
- › Look for shared segments that are long enough to be highly unlikely to be shared by chance



Relatedness Using SNPs

Amount of shared DNA (in cM) across the genome correlates with relatedness

Closer relatives share more DNA



Ball, et al.
*Ancestry
DNA
Matching
(2016)*

Edited by
Leah Larkin,
Ph.D.

Genetic Genealogy Databases

- › Law enforcement usage allowed for unidentified remains and violent crimes
 - GEDmatch
 - FamilyTreeDNA (FTDNA)
- › No law enforcement usage
 - 23andMe
 - AncestryDNA
 - MyHeritage

Database Searching

- › Compare queried file to all opted-in users' files
- › Only the amount and location of shared DNA is shown, **not the DNA itself**

Chr	Start Location	End Location	Centimorgans (cM)	SNPs
3	14,335,343	26,999,951	15.9	2,036
6	165,997,807	170,693,361	11.2	1,018
21	25,436,468	46,897,344	44.4	4,233

Largest segment = 44.4 cM

Total of segments > 7 cM = 71.5 cM

3 matching segments

Estimated number of generations to MRCA = 3.8

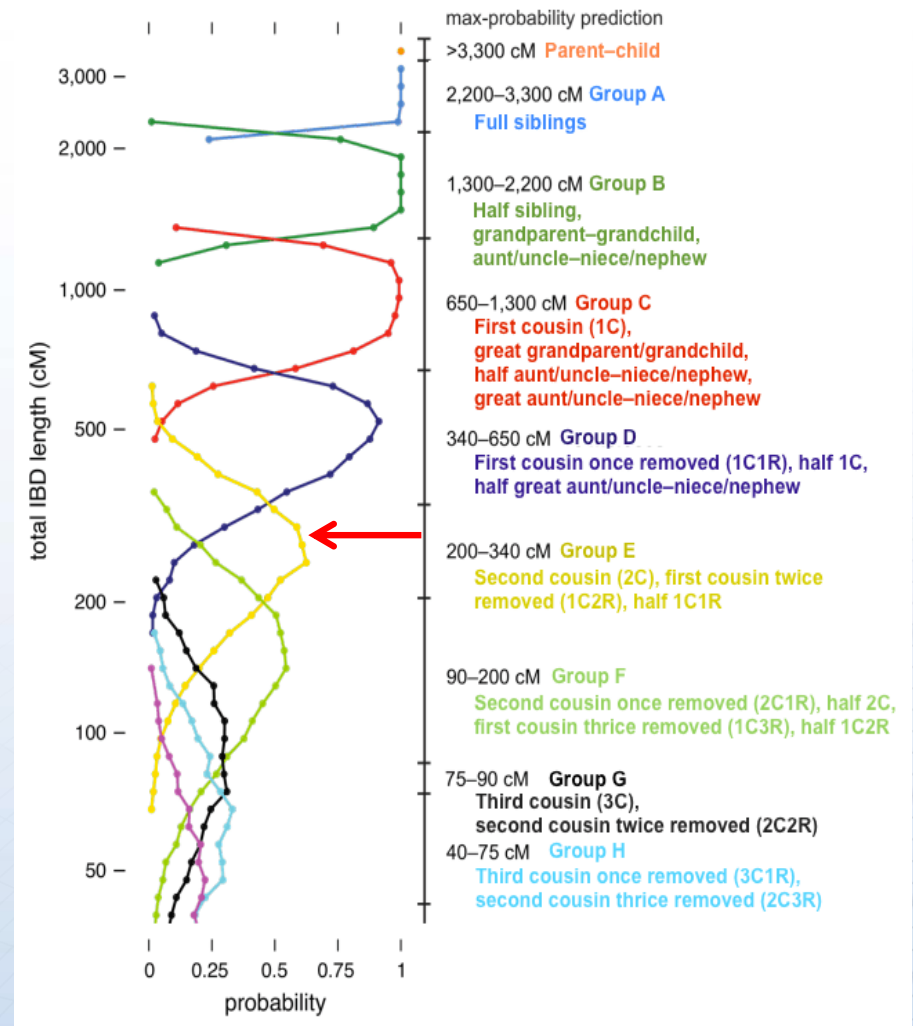
406488 SNPs used for this comparison.

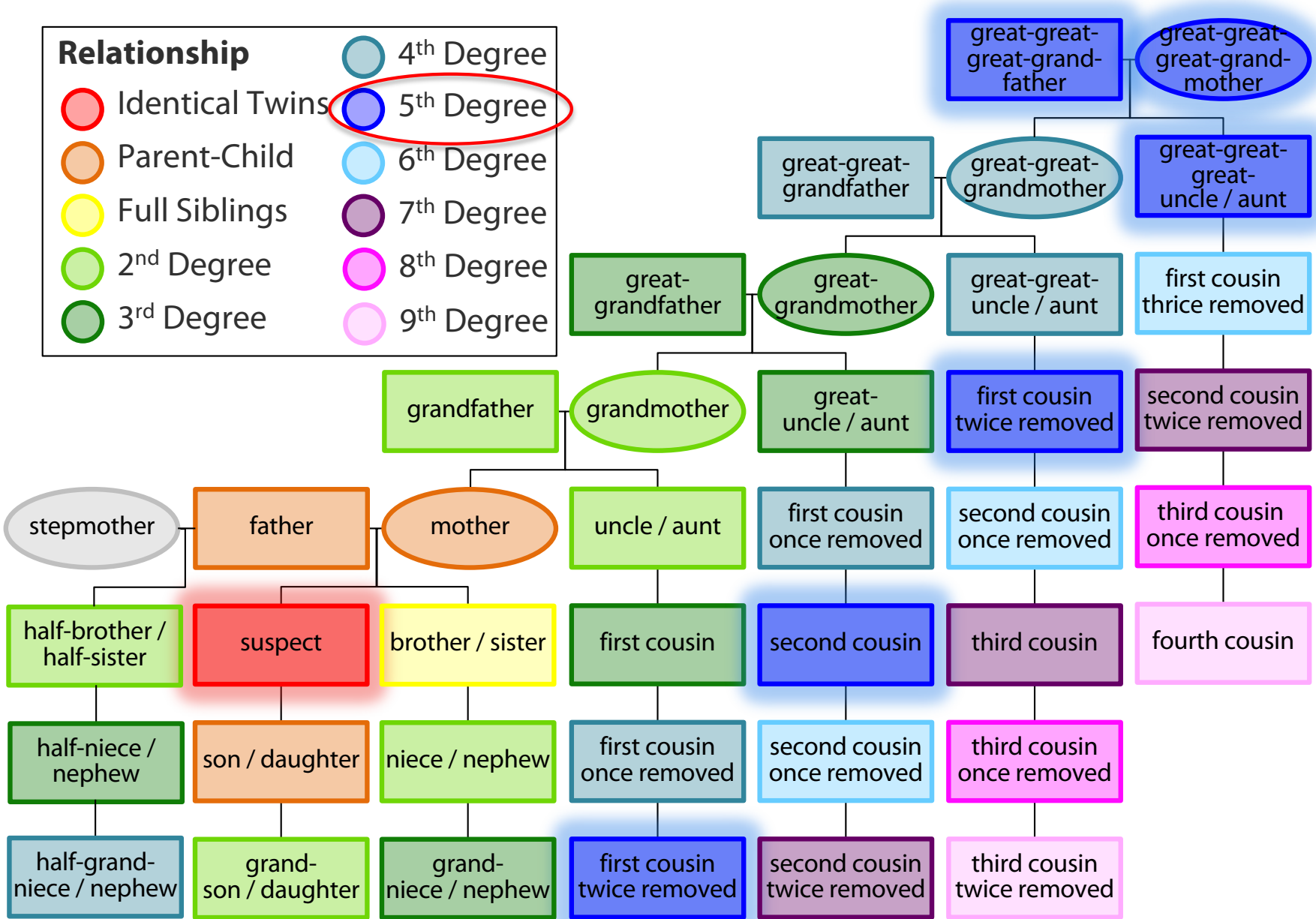
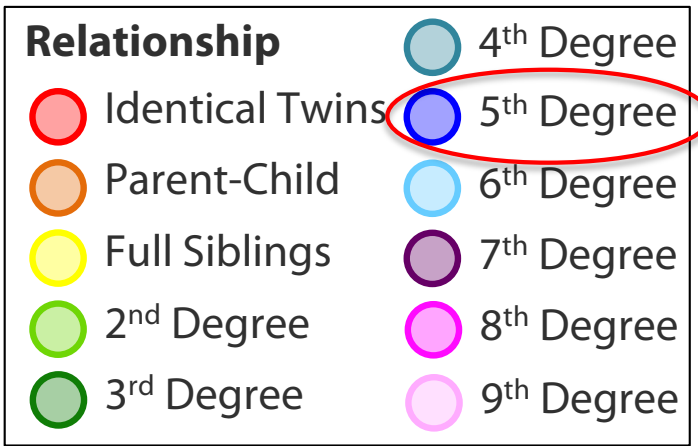
Procedure

1. Generate DNA data from a forensic sample from an unknown subject
2. Upload DNA data to a database to find other individuals that share DNA with the subject
3. Build family trees back in time to possible common ancestors using public records

Genetic Genealogy

- › Total shared DNA: 300 cM
- › The amount of sharing can be consistent with multiple possible degrees of relatedness
- › Each degree of relatedness has multiple possible relationship types



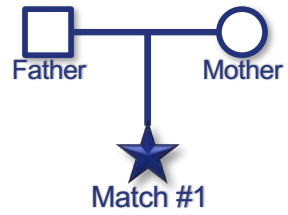


Family Tree Building

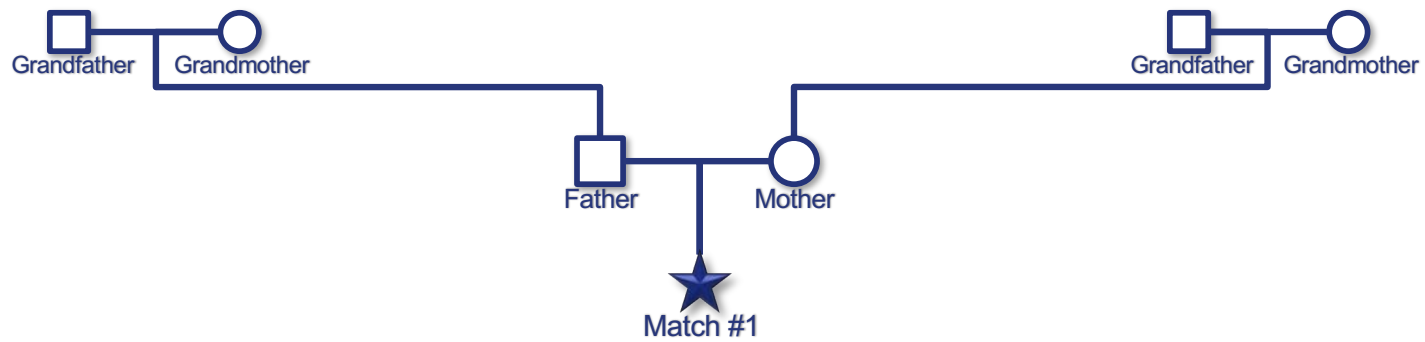


Match #1

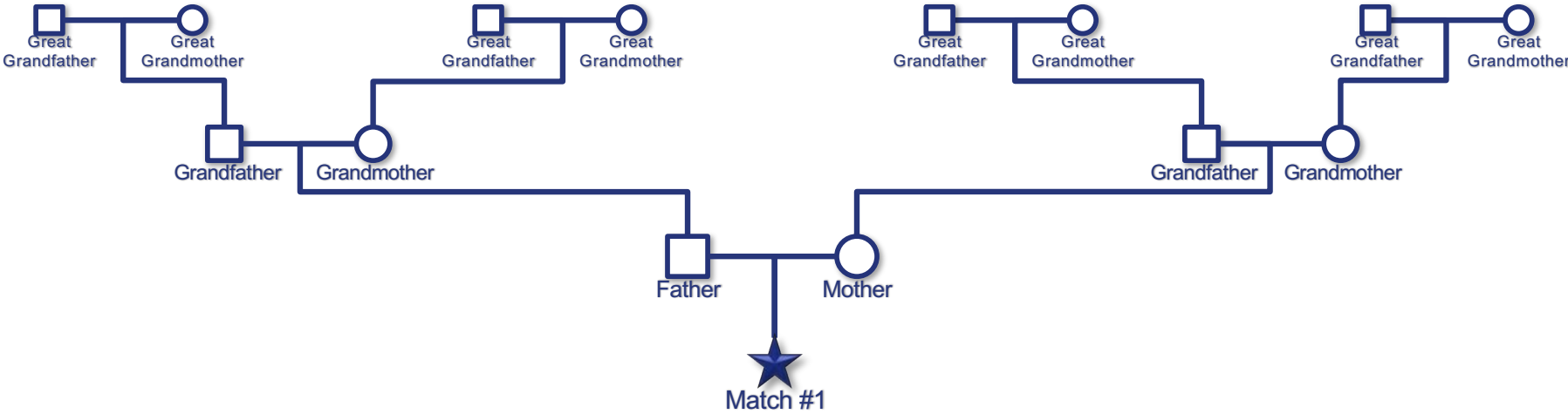
Family Tree Building



Family Tree Building



Family Tree Building



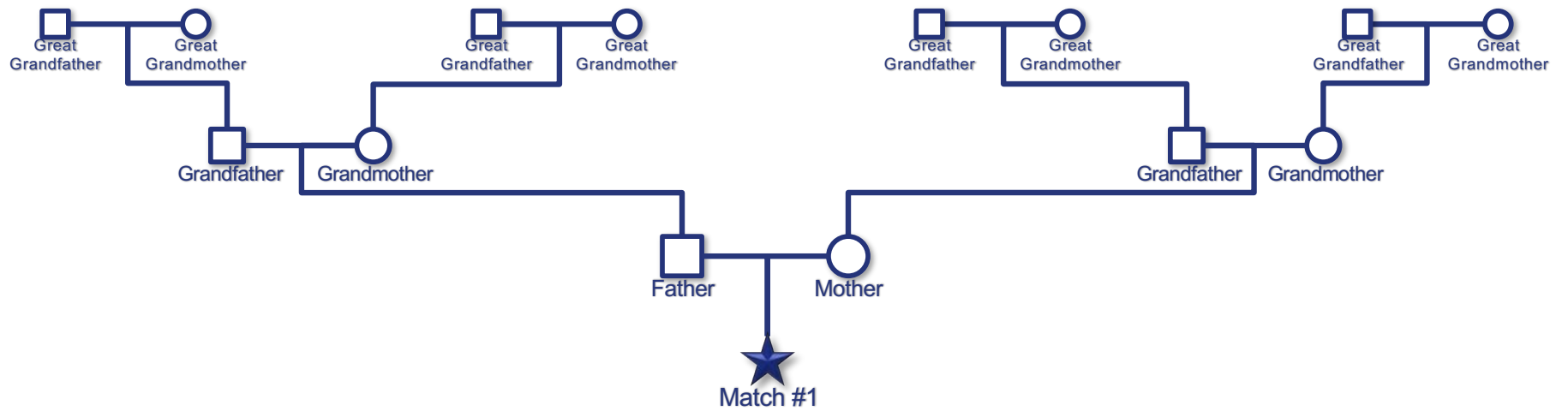
Family Tree Building

- › Records are not always available
- › Biological family trees don't always match family trees on paper
 - Misattributed paternity
 - Unknown parentage
 - Adoption

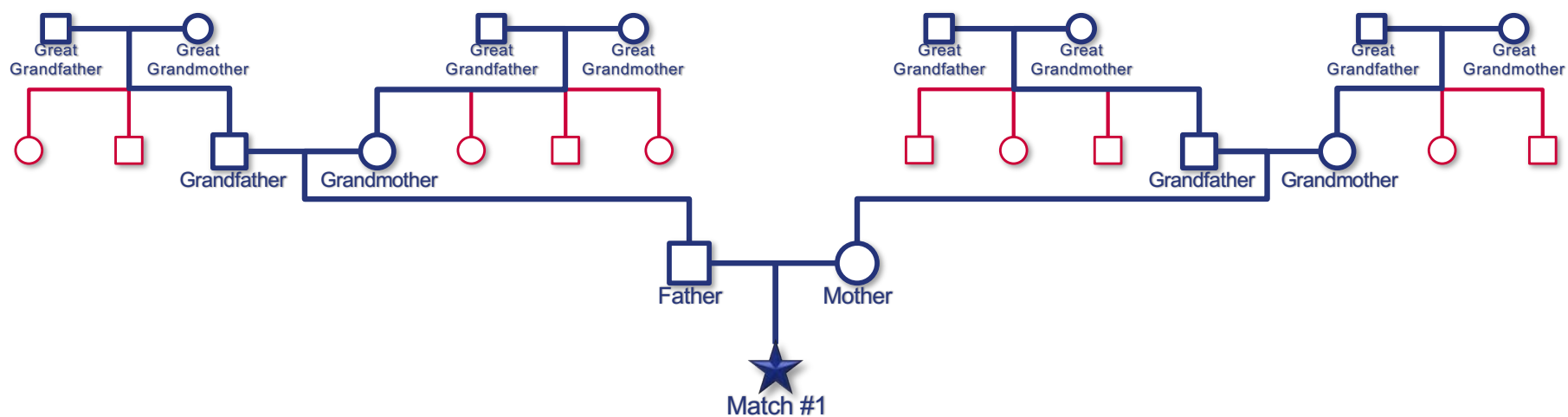
Procedure

1. Generate DNA data from a forensic sample from an unknown subject
2. Upload DNA data to a database to find other individuals that share DNA with the subject
3. Build family trees back in time
4. Identify all of the other descendants of those common ancestors at the right genetic distance to be the Subject

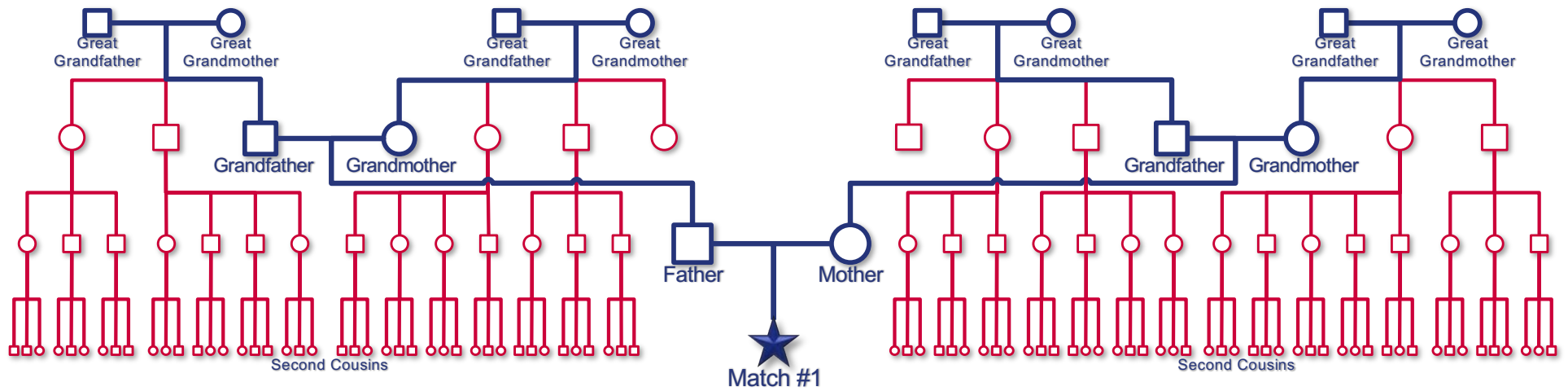
Descendancy Research



Descendancy Research



Descendancy Research



Descendancy Research

- › Not all descendants can be identified
 - Out of wedlock births
 - Misattributed parentage
 - Families can be very large

Procedure

1. Generate DNA data from a forensic sample from an unknown subject
2. Upload DNA data to a database to find other individuals that share DNA with the subject
3. Build family trees back in time
4. Identify all of the other descendants of those common ancestors at the right genetic distance
5. Narrow down the possible identities

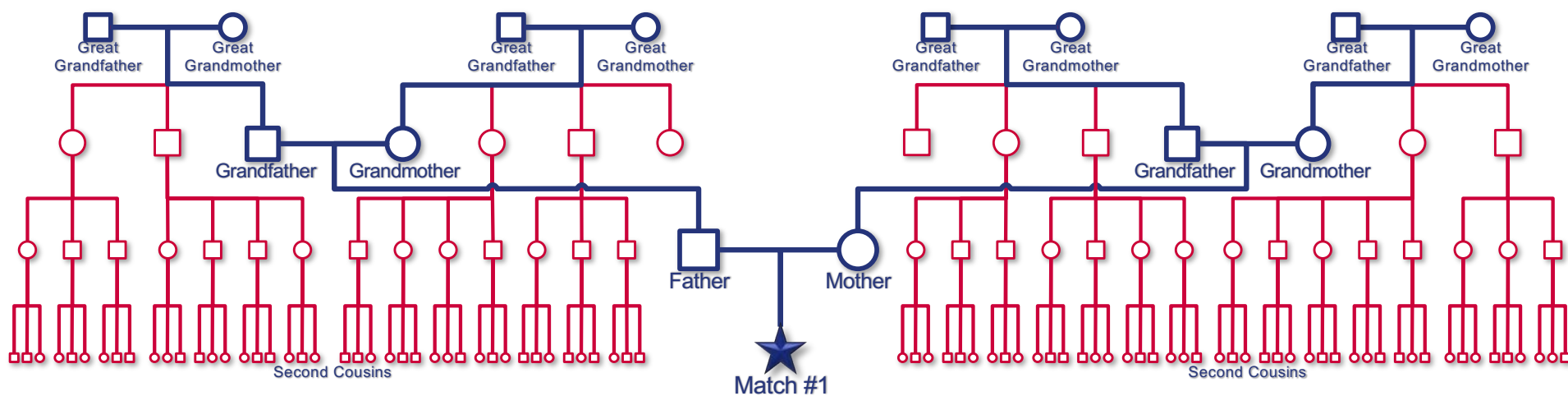
Narrowing the search

- › Demographic criteria
 - Sex of the subject
 - Date of the crime
 - Location of the crime
- › Genetic sharing between matches (triangulation)
- › Ancestry and Phenotype predictions
- › Targeted Kinship testing

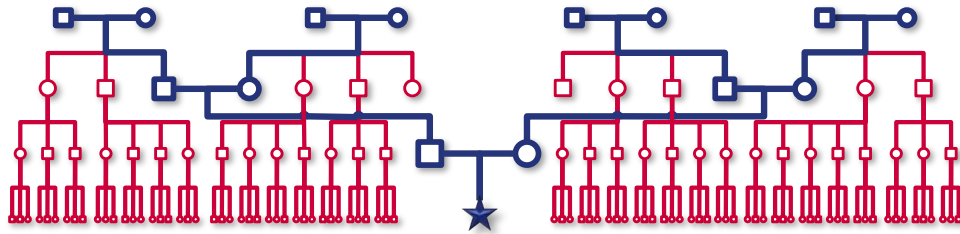
Triangulation

- › If there are multiple matches to the subject that don't share DNA with each other, they represent different branches of the subject's family tree
- › The goal is to find an intersection between their family trees (a marriage or out-of-wedlock birth)

Triangulation

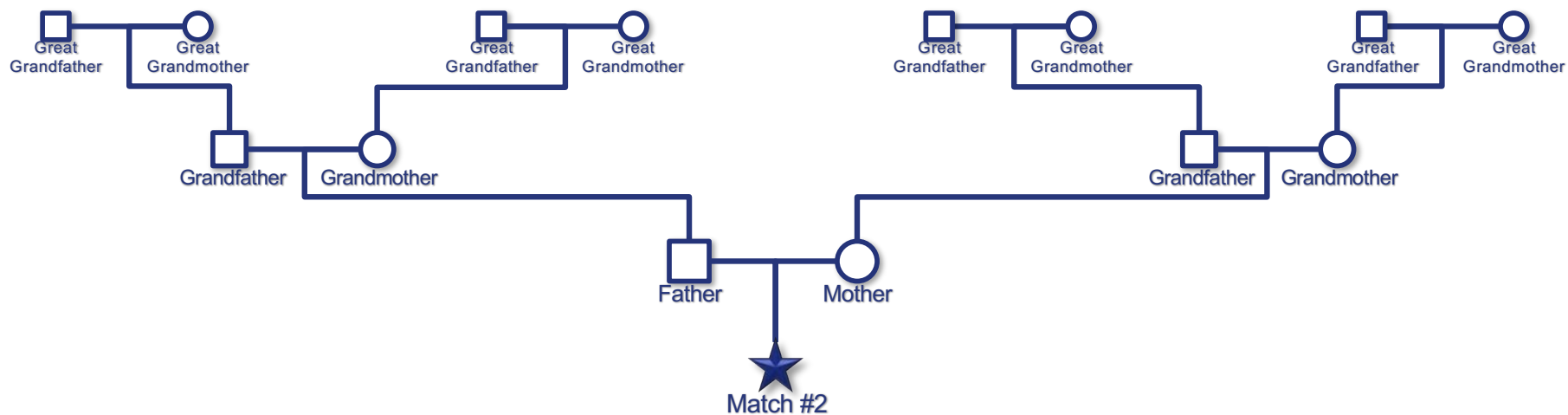
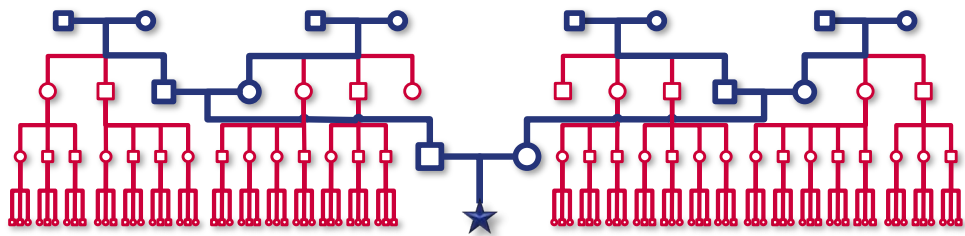


Triangulation

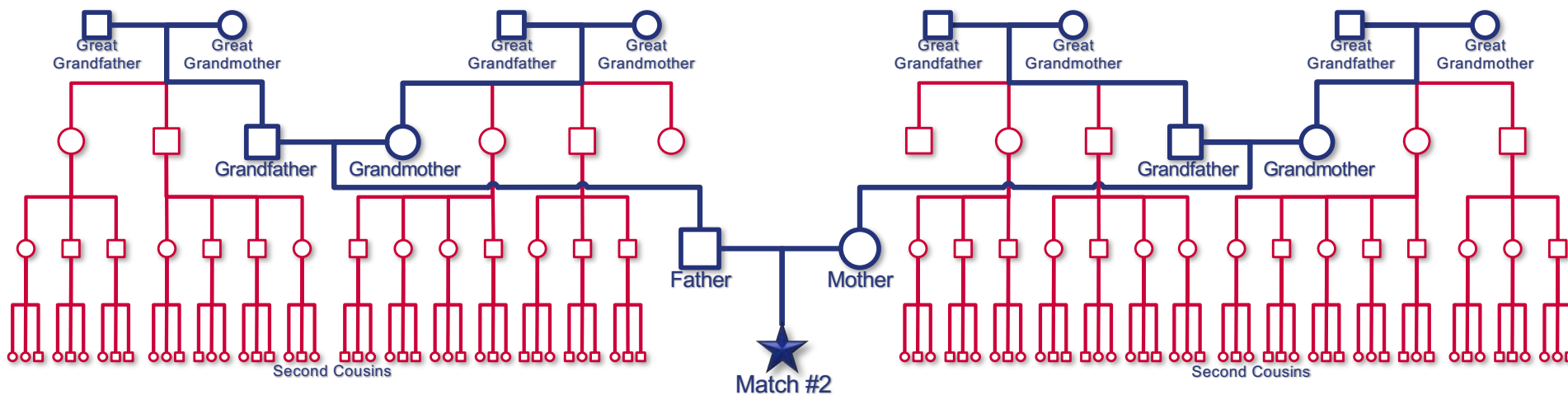
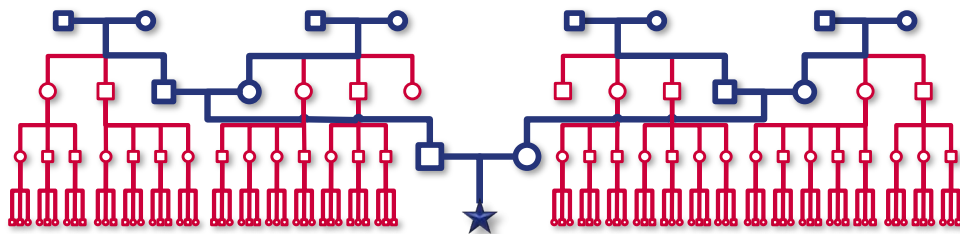


★
Match #2

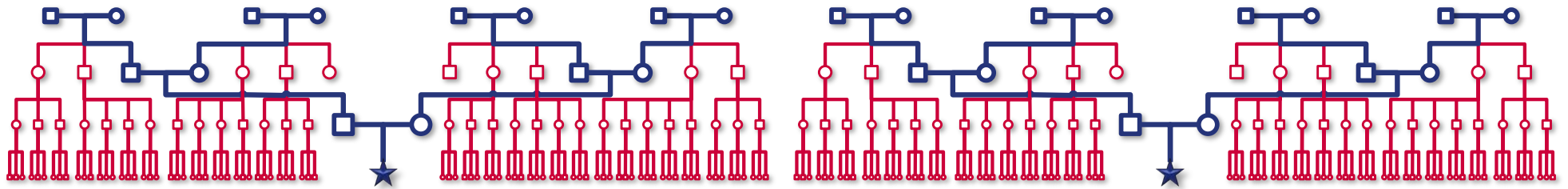
Triangulation



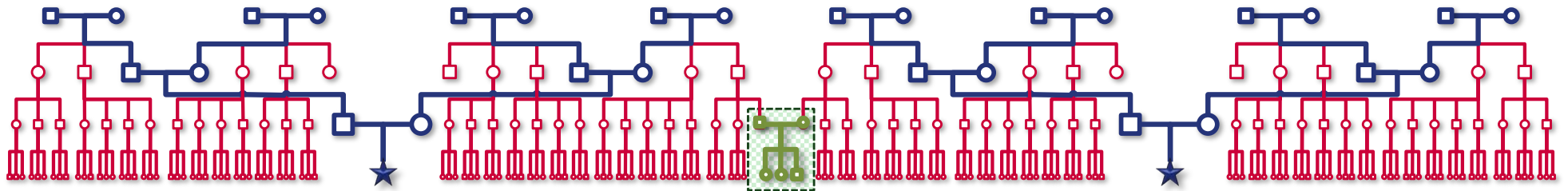
Triangulation



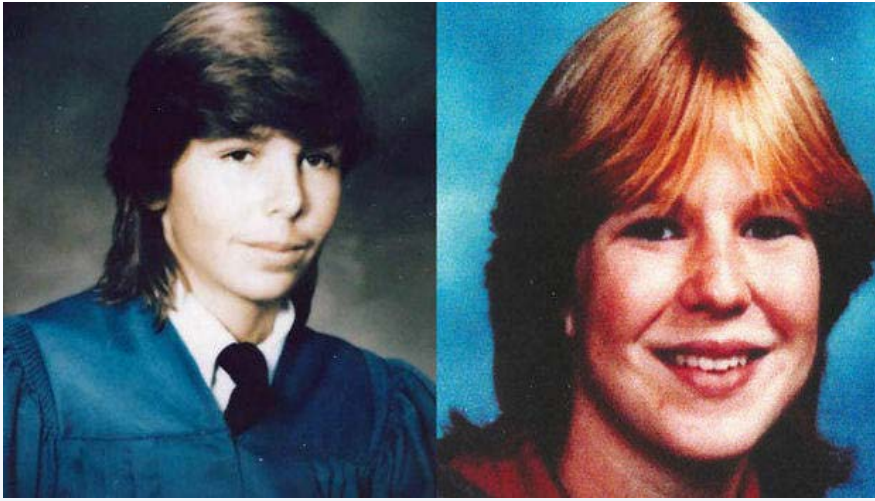
Triangulation



Triangulation



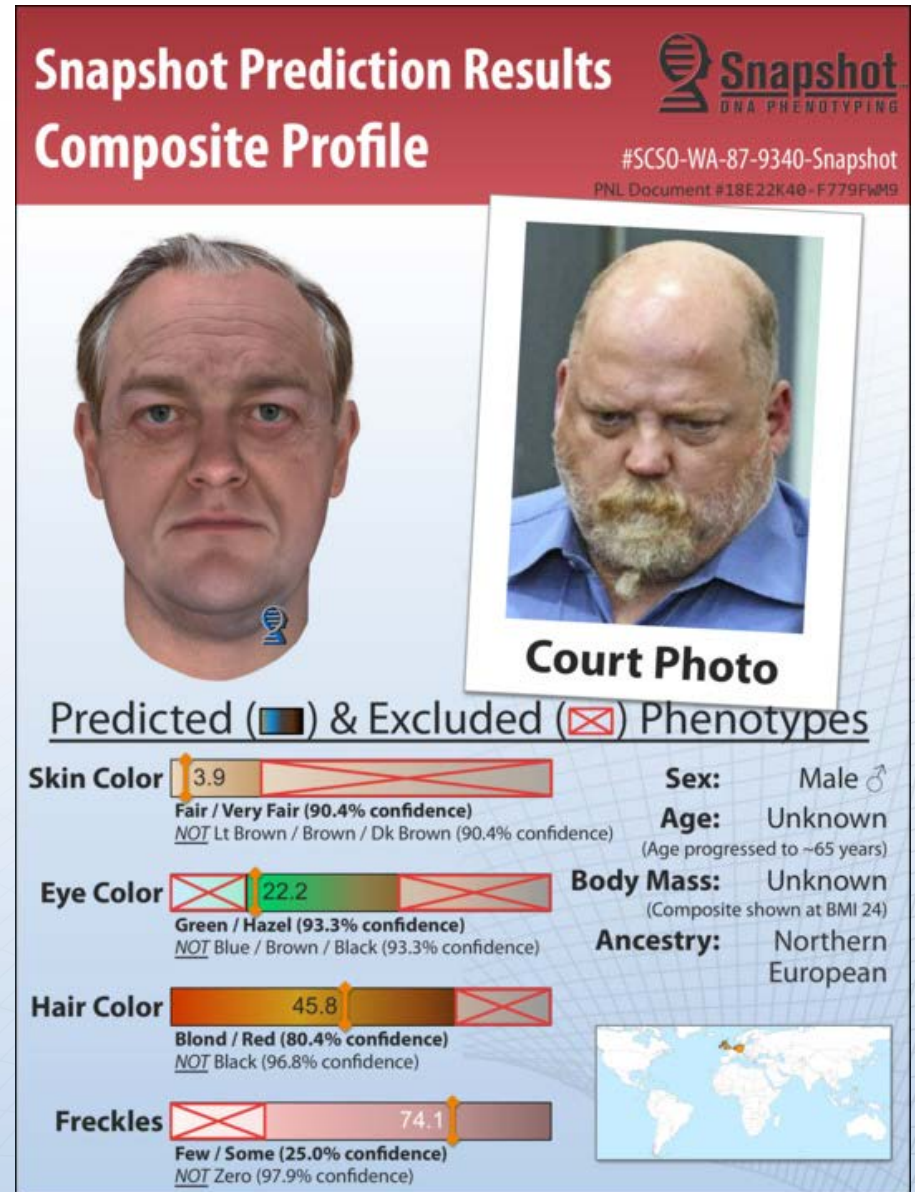
Case Study #1



Jay Cook

Tanya Van Cuylenborg

- › 1987 double murder of Jay Cook (20) and Tanya Van Cuylenborg (18)
- › Snohomish County, WA



Case Study #1

Cook/Van Cuylenborg Double Homicide Cold Case



Suspect family tree based on genetic genealogy



MATCH
Cousin?
#1

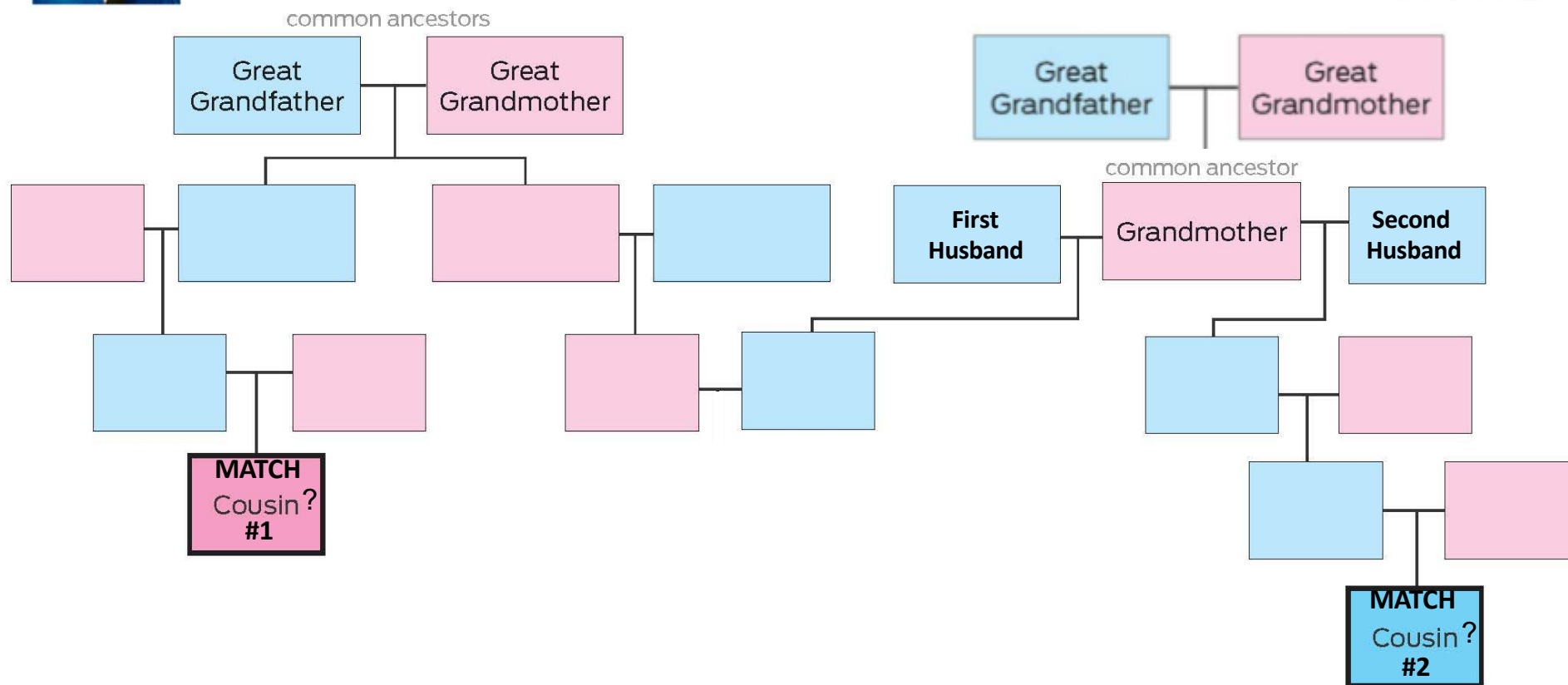
MATCH
Cousin?
#2

Case Study #1

Cook/Van Cuylenborg Double Homicide Cold Case



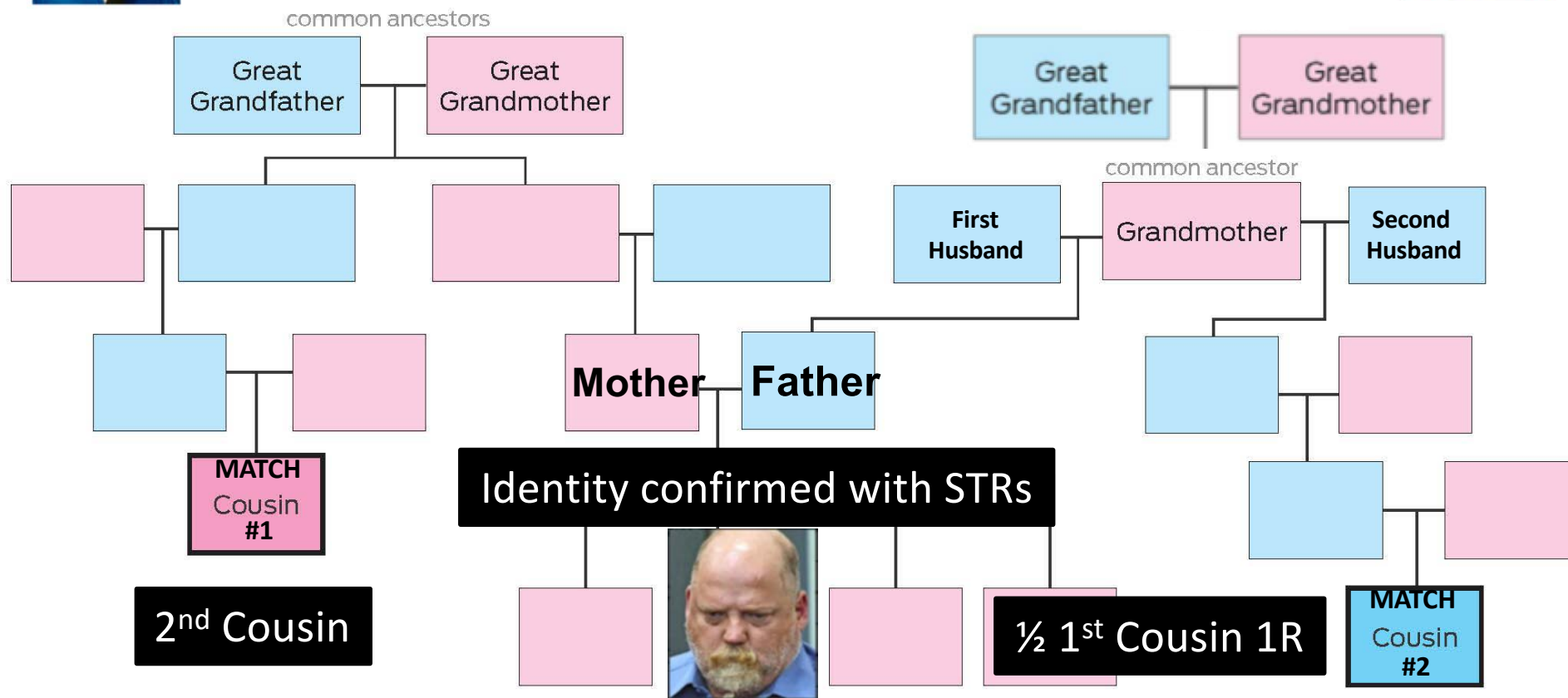
Suspect family tree based on genetic genealogy



Case Study #1

Cook/Van Cuylenborg Double Homicide Cold Case

Suspect family tree based on genetic genealogy

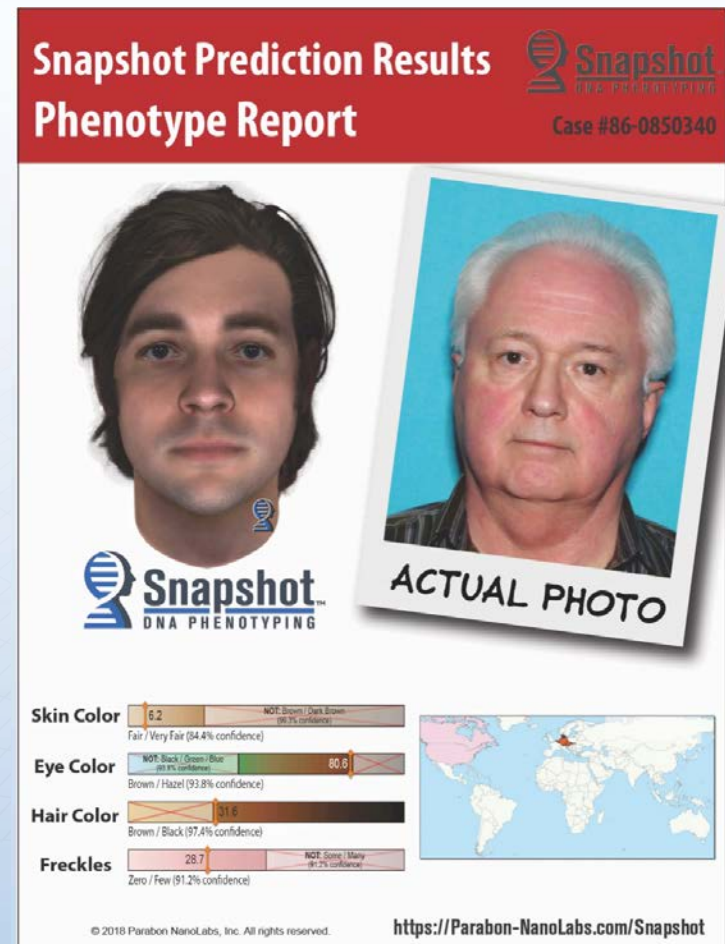


Case Study #2

- › 1986 homicide of 12-year-old Michella Welch
- › Tacoma, Washington



Michella Welch, 12

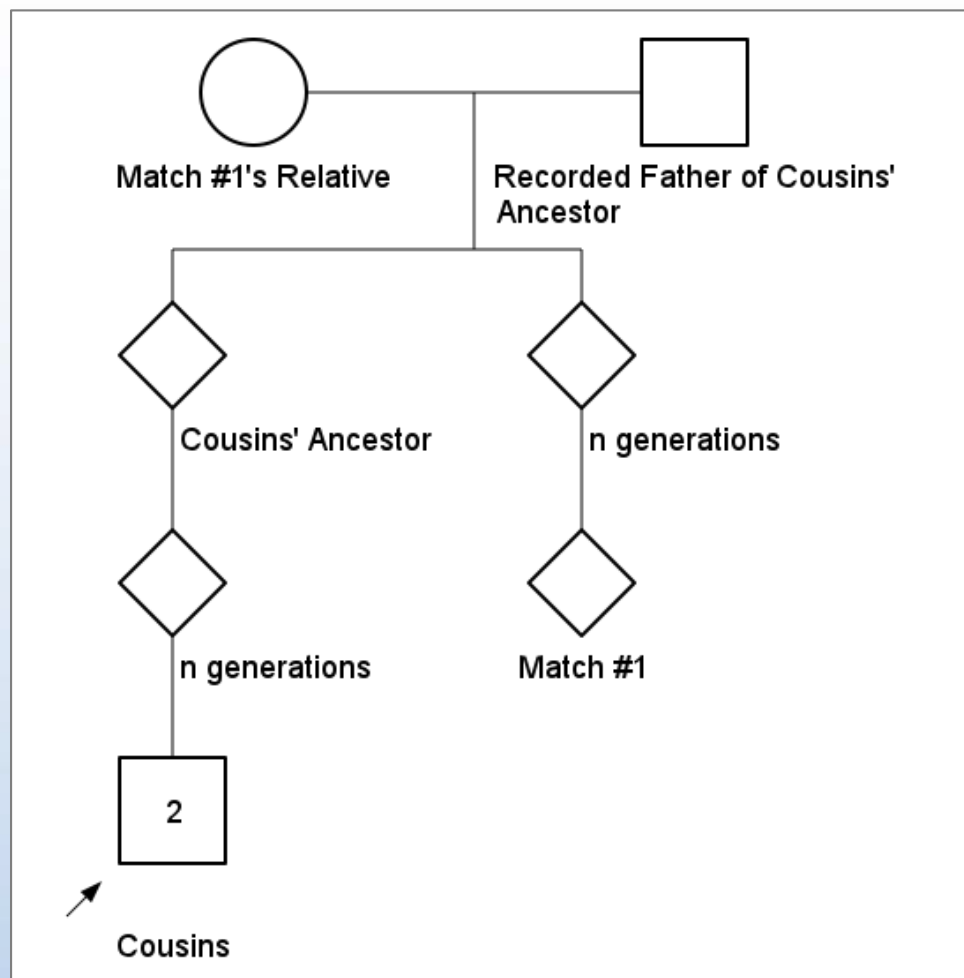


Case Study #2

- › Predicted genetic ancestry: 90% Northern European + 10% North Native American
- › Top two matches in GEDmatch shared no DNA
- › Pair of brothers of possible interest
 - Cousins to Match #1
 - Lived in Tacoma in 1986
 - Had two Native American great-great-grandmothers (~1/8 Native American)
 - But ...

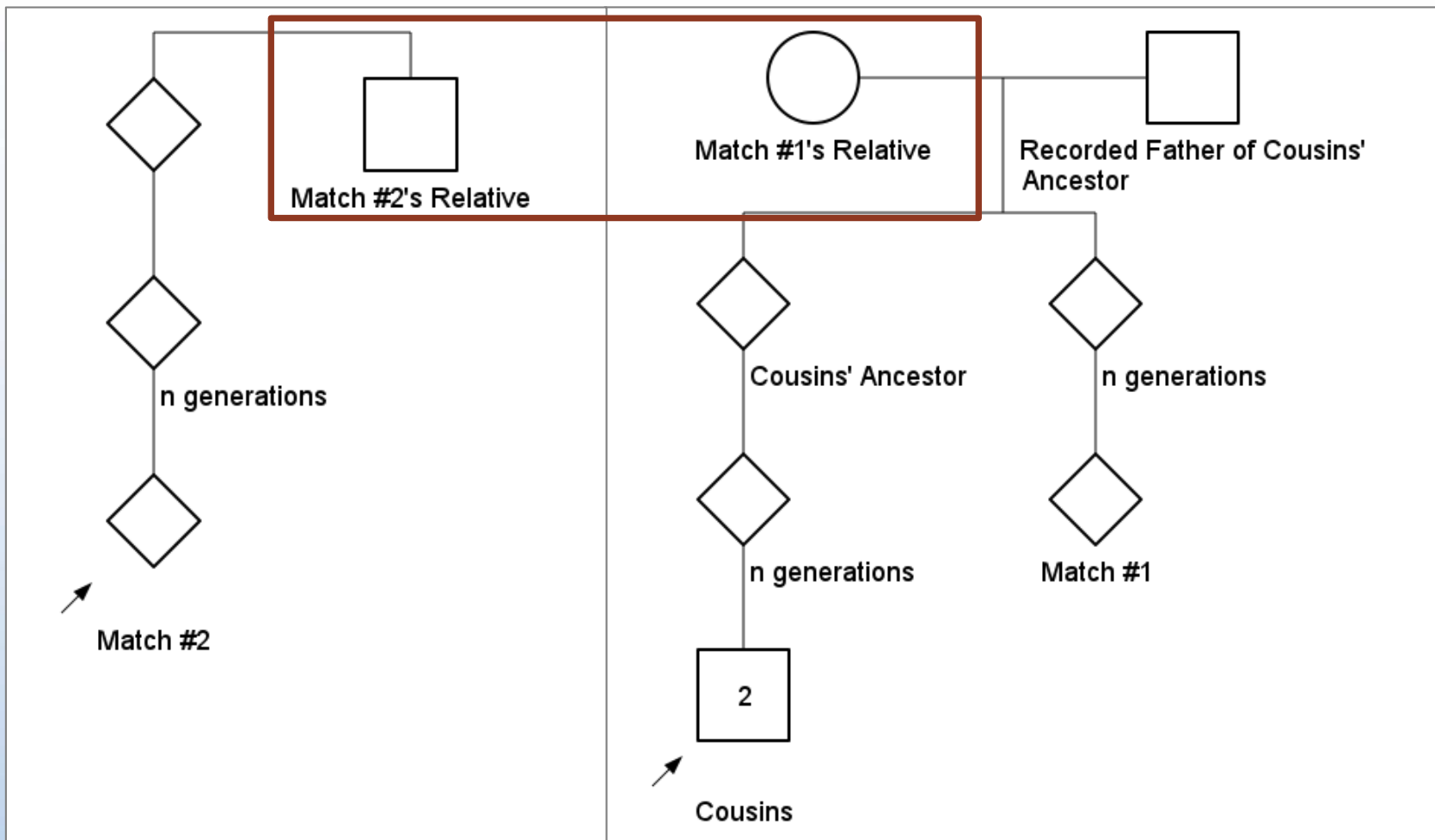
Case Study #2

- One degree closer than expected given amount of shared DNA; no connection to match #2



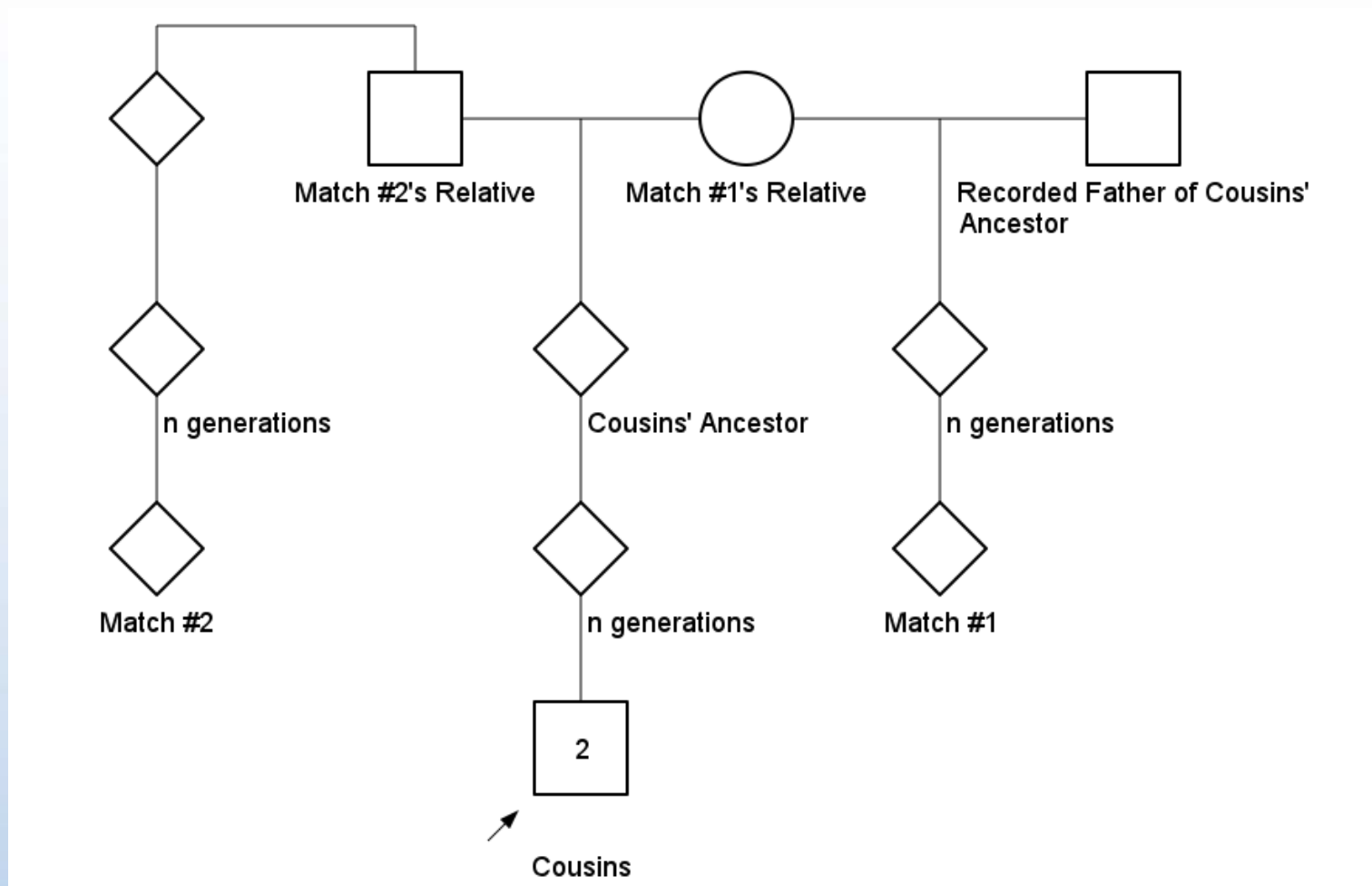
Case Study #2

- Match #1's Relative and Match #2's Relative lived in the same small town in the year before the Cousins' Ancestor was born



Case Study #2

- Misattributed paternity can account for the amount of shared DNA and the connection to Match #2



Forensic Analysis of Next-Generation Sequencing (NGS) Data

PARABON[®] F_x[™]
Forensic Analysis Platform



Forensic NGS

- › NGS is a powerful new tool for forensics
 - STRs and SNPs in a single assay
 - Sequencing STRs yields additional alleles that are helpful for answering identity questions and analyzing mixed samples

Forensic NGS

- › **DoD problem:** multiple NGS system manufacturers, each with their own proprietary software
 - Can't compare samples across NGS kits
 - Analytical capabilities are limited to those tools included in manufacturers' software
- › **Solicitation:** platform-agnostic software that can analyze STRs, SNPs, and mitochondrial DNA from NGS data

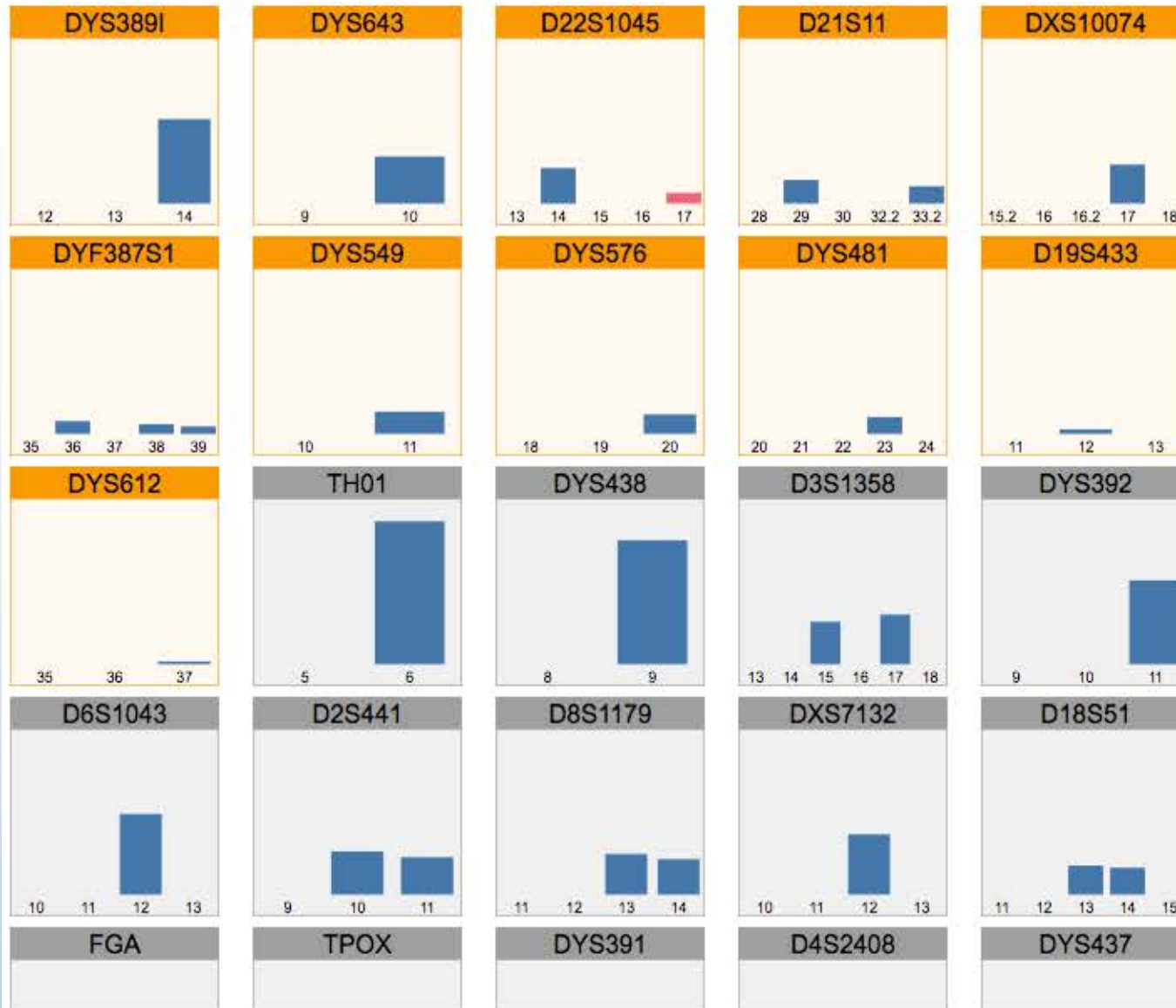
Forensic NGS

› Parabon Fx™ Forensic Analysis Platform

- Analyze raw data from any NGS system and compare across kit types
- Extensible platform with a plugin architecture
- Dozens of built-in plugins provide extensive analytical capabilities
- Can be installed on a local server or accessed over the internet
- CAC card access, STIG-compliant

NGS STRs

- › Create, edit, and approve an STR profile using allele length or sequence



NGS STRs

> Deconvolute mixed STR profiles using length or sequence alleles

Mixture Deconvolution Analysis Results

Weight of Evidence

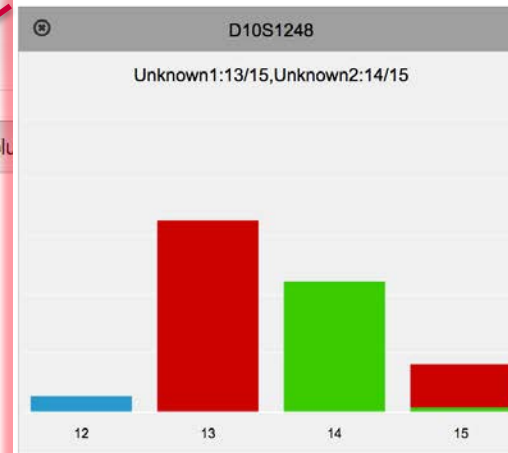
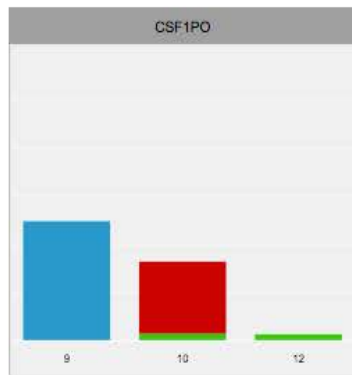
MLE Parameter Estimates

Deconvolution Hd

Deconvolu

Profile View

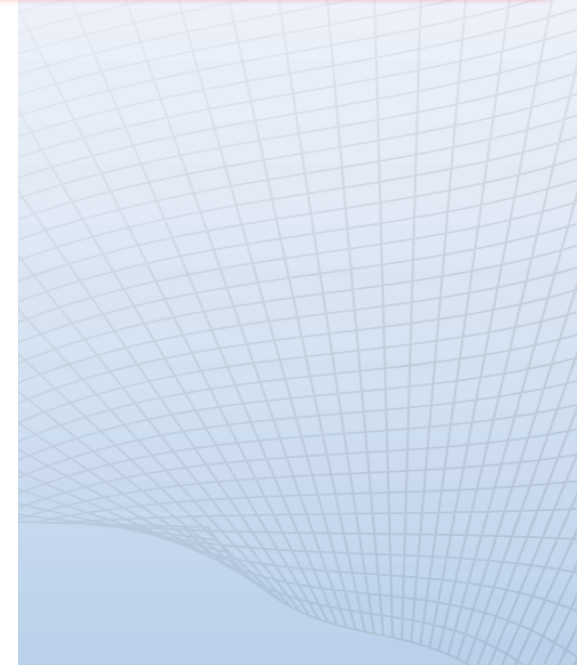
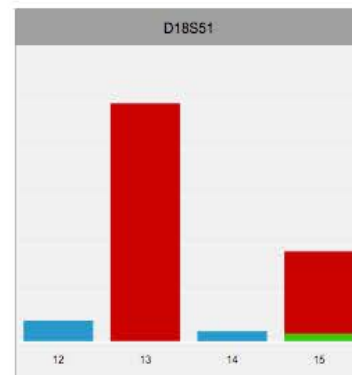
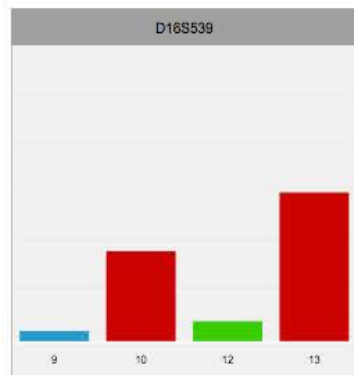
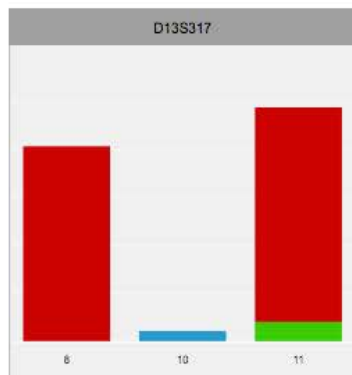
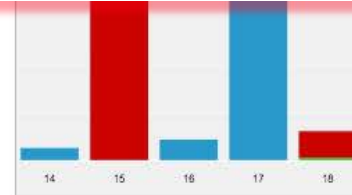
List View



D10S1248:
Unknown1:13/15,Unknown2:14/15

Allele	ReadCount	Contributor
13	326	Unknown1
14	222	Unknown2
15	73.452	Unknown1
15	6.548	Unknown2
12	26	Unassigned

Unknown2
Unknown1
Unassigned



NGS SNPs

- › Align sequence reads to the reference genome and call a SNP profile



NGS SNPs

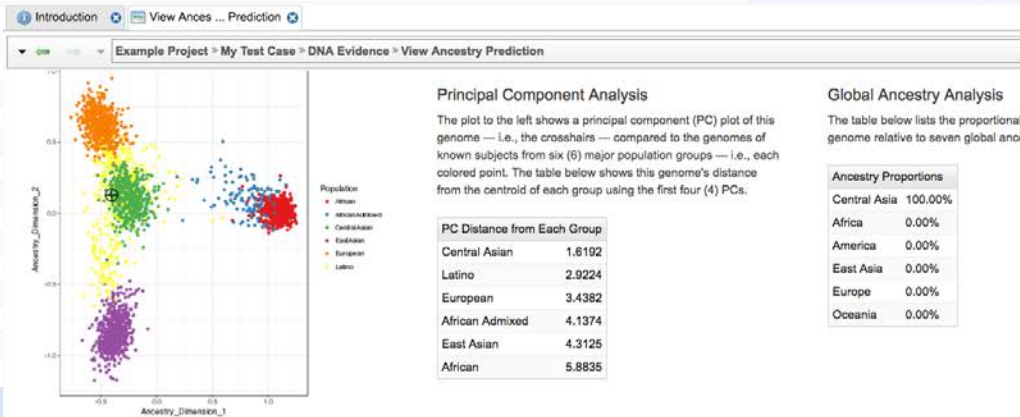
› Perform identity matching, infer ancestry, predict phenotype

🔍 Sequence Data SNP Identity Match Results

Project Name	Case Name (Reference)	Sample Name (Reference)	Data Set
Example Project	My Test Case (19801208-NY-JWL)	DNA Evidence	#8323: Example Project » My Test Case (19801208-NY-JWL) » DNA Evidence » SNP Ident

This table summarizes the SNP profiles of the selected source and target samples. At each locus, each individual's alleles are shown, along with whether the locus is a full match, partial match, or mismatch. Mismatching loci are shown at the top. Click on a row to see plots of the read counts for each sample at that locus.

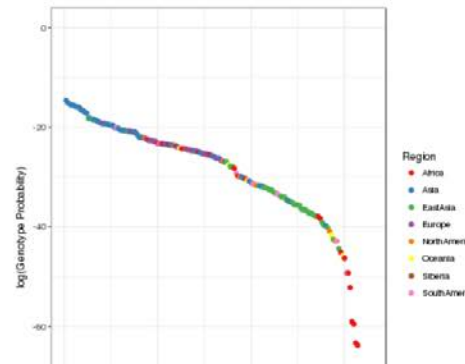
Locus	DNA Evidence	Bart Baker Sr.	Match Type
rs10776839	T	G	Mismatch
rs214955	T	C,T	Partial
rs279844	A,T	A	Partial
rs354439	A	A,T	Partial
rs430046	C	C,T	Partial
rs445251	C,G	G	Partial
rs729172	G,T		
rs826472	C,T		
rs873196	T		
rs891700	A,G		
rs901398	C		
rs907100	C		
rs964681	C,T		
rs993934	A,G		
rs1024116	C,T		
rs1028528	A,G		
rs1031825	C		
rs1058083	G		
rs1109037	G		
rs1357617	T		
rs1454361	A,T		



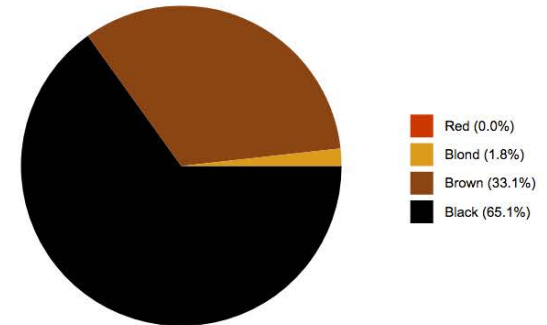
Population Likelihood Analysis

The table below shows this genome's 10 most likely populations from FrogKB. The plot to the right shows the Log Probability of each population, colored according to region.

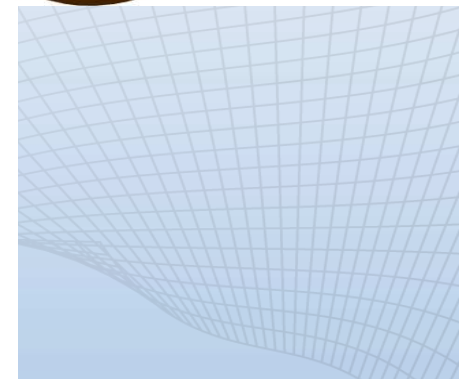
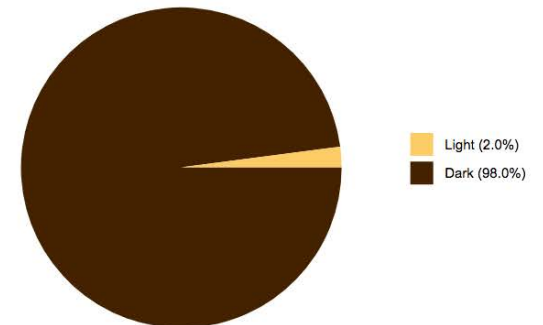
Population	Region	Genotype Probability	Likelihood Ratio	Log Probability
Kachari	Asia	2.4545e-15	1.0000	-14.6100
Gujarati(GIH)	Asia	8.6251e-16	2.8458	-15.0642
Sri_Lankan_Tamil(STU)	Asia	4.0454e-16	6.0875	-15.3930
Keralite	Asia	2.7526e-16	8.9165	-15.5002
Bengali(DEB)	Asia	2.5915e-16	9.4716	-15.5865
Punjabi(PJL)	Asia	1.5435e-16	15.9021	-15.8115
Telugu(ITU)	Asia	1.3321e-16	18.4262	-15.8755
Tholi	Asia	9.4926e-17	25.8576	-16.0226
Mohannas	Asia	2.8118e-17	87.2938	-16.5510
Pathan	Asia	2.2810e-17	107.6063	-16.6419



Hair Color: Black

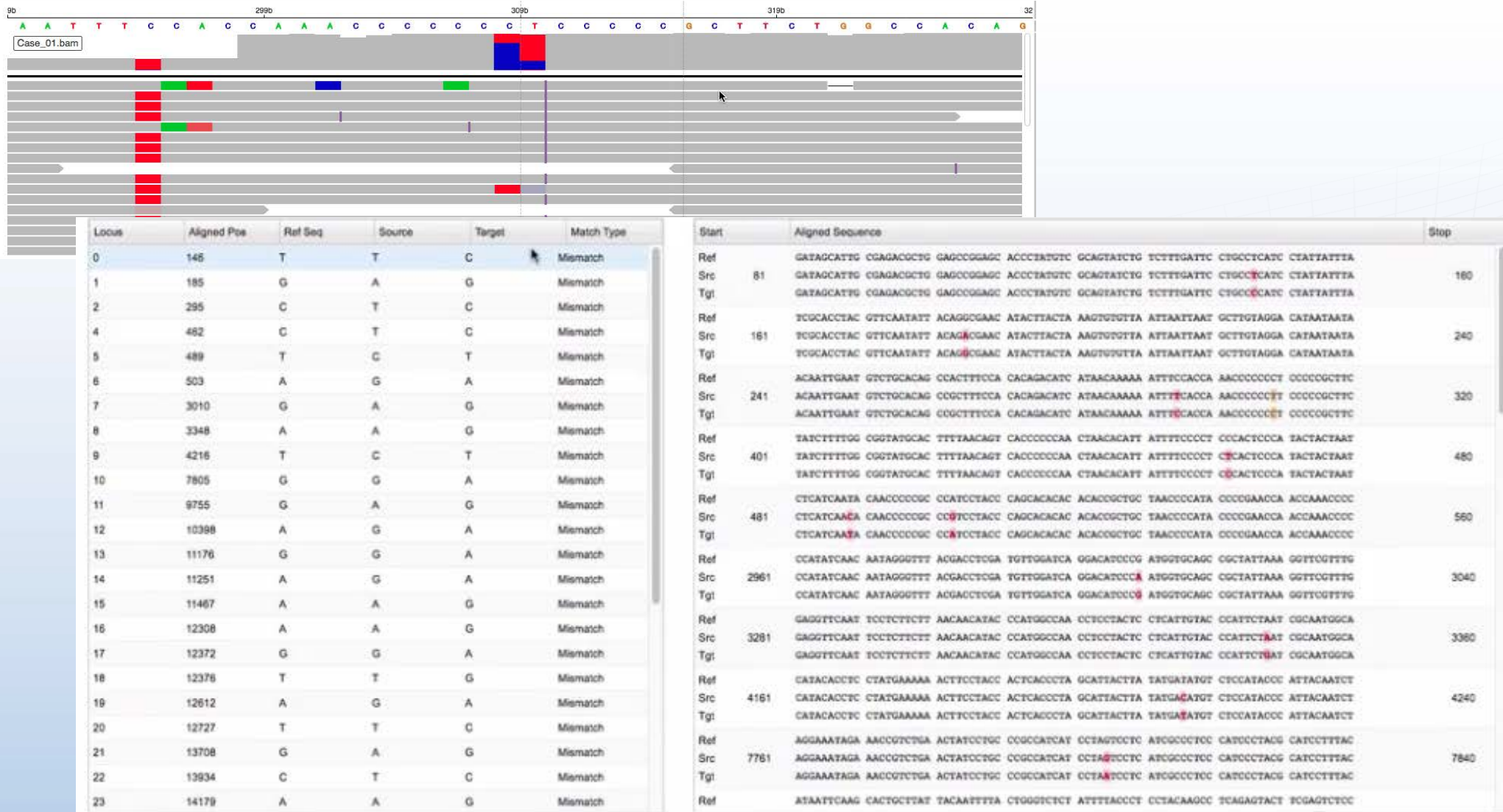


Hair Darkness



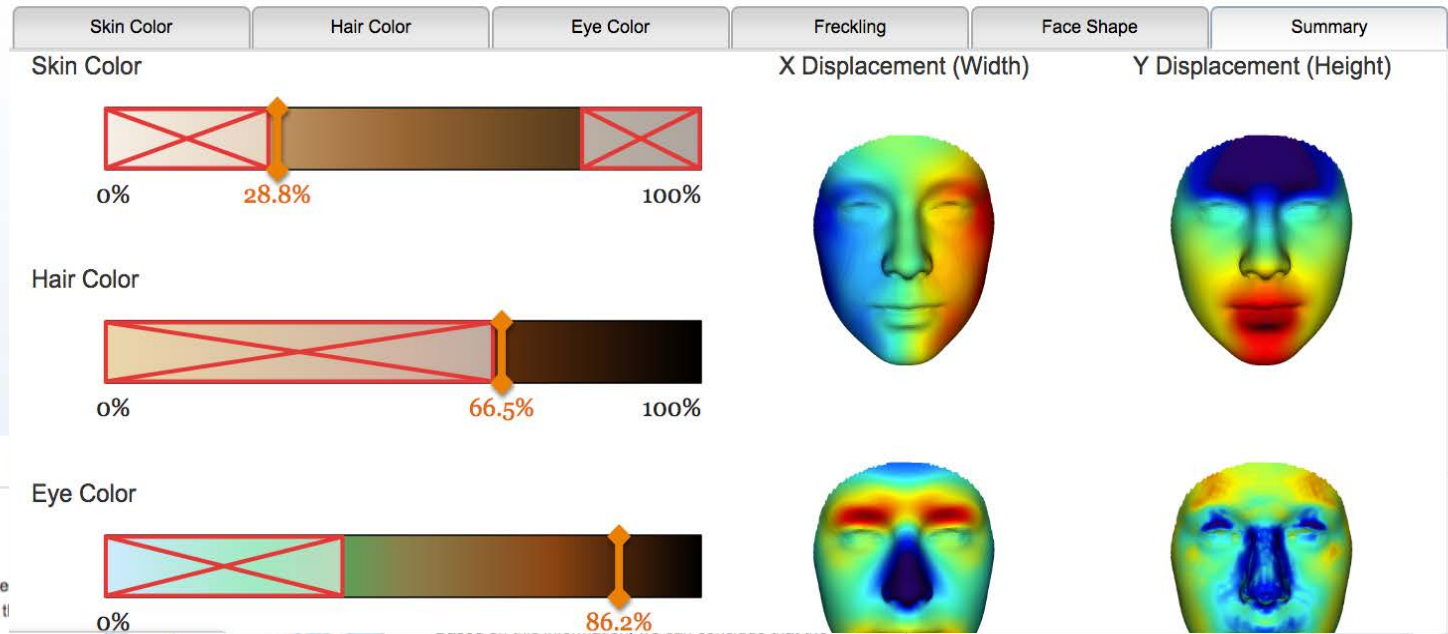
NGS mtDNA

- Create a mitochondrial DNA sequence profile and perform haplotype-based or string-based lineage matching



Microarray Plugins

- > The DoD distribution of Fx also includes plugins for ancestry, phenotype, and kinship analysis of SNP microarray data using Snapshot



Regional Ancestry Analysis

The table below lists the proportional membership of this genome relative to 32 regional populations, which are illustrated on the map to the right.

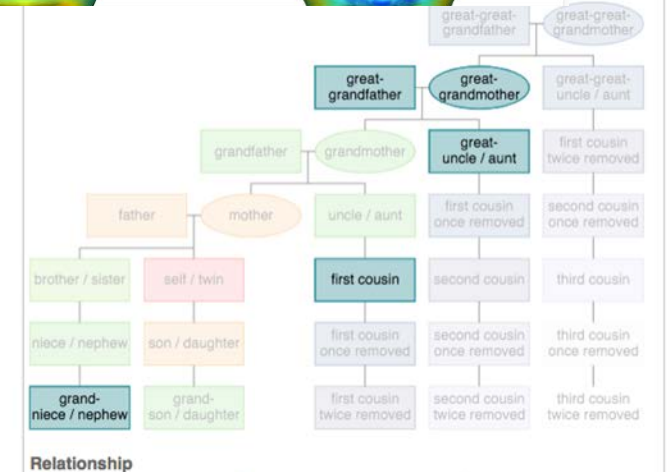
Region	Percent
Europe--Southwest	43.00%
Americas--Central	28.91%
Americas--South	16.58%
Middle East--Arabia	8.63%
Other	2.88%

Only population groups comprising at least 5% of the genome's ancestry are listed; population groups comprising less than that have been combined into "Other" (not shown on map).



Individuals represented by these samples are related with a confidence of 0.99470. Furthermore, we can conclude that they have a relationship of 3rd Degree with a confidence of 0.91868.

Outcome	Probability
3rd Degree	0.98534
4th Degree	0.01224
6th Degree	0.00065
5th Degree	0.00052
Identical Twins/Self	0.00031
2nd Degree	0.00027
Unrelated	0.00024
Full Siblings	0.00023
Parent Offspring	0.00020



Fx Extensions

- › New analyses developed by Parabon can quickly be distributed to DoD users as Fx plugins
 - Kinship analysis of highly degraded DNA
 - Ancestry analysis of highly degraded DNA
 - Automated pedigree reconstruction
 - Automated mixture deconvolution
 - Etc.

Questions?

ellen@parabon.com

