

Aspects of Phenotyping and Ancestry – and How Should We Report?



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KING'S
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LONDON

Outline – Phenotypics in King's Forensics

Age

- Current capabilities and continuing research

Skin colour

- Applying the HIrisPlex-S snps for skin colour

Biogeographical ancestry

- Use in the UK population
- Using sequence variants

Reporting

- Different approaches

Conclusions

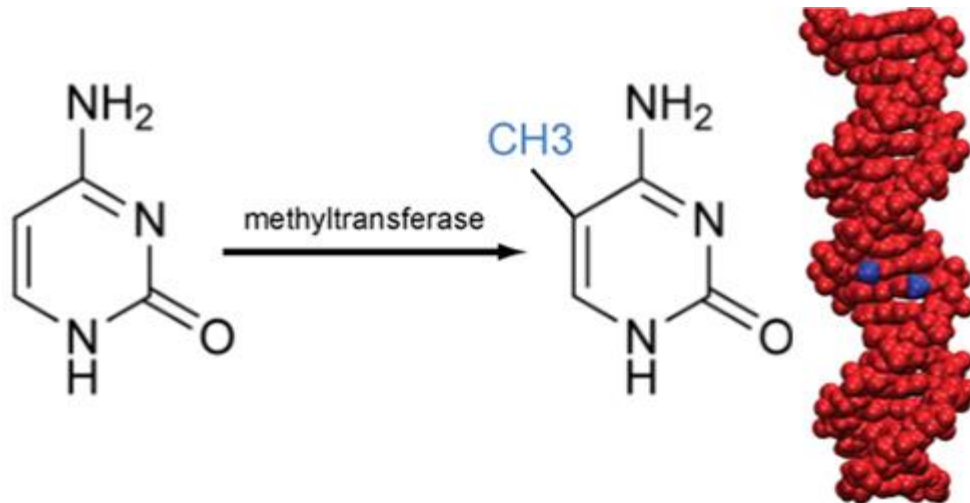
AGE – epigenetic tools for blood stains

Anastasia Aliferi

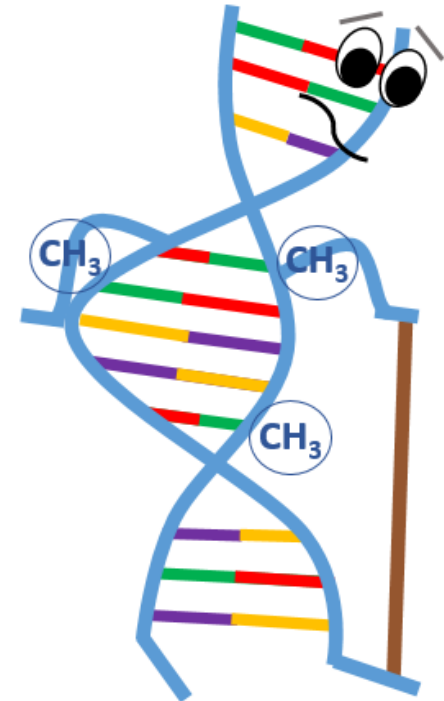
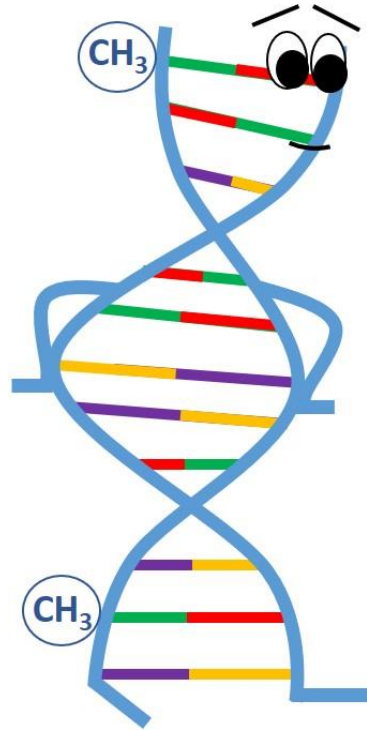
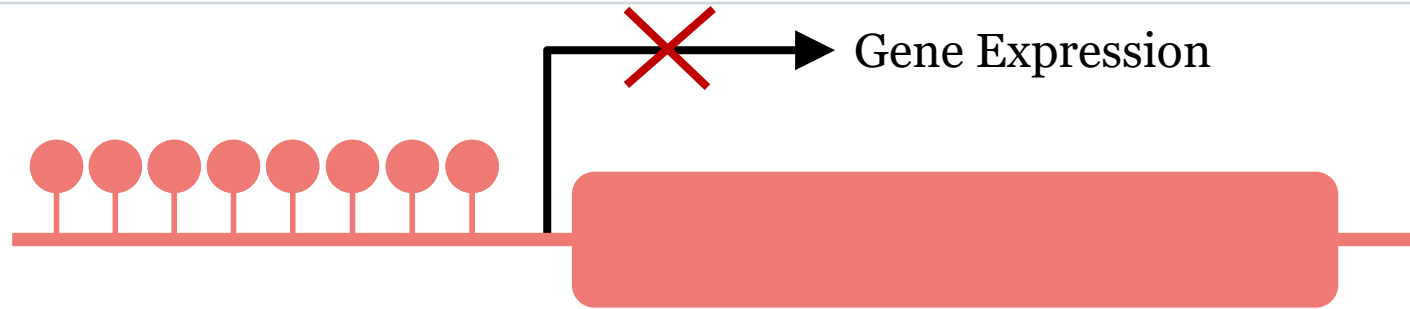


Age prediction – different strategies

- Telomere shortening
- Genetic rearrangements
- Epigenetic factors
 - Histone modifications
 - Nucleosomal remodelling
 - Chromatin looping
 - Non coding RNAs
 - DNA methylation



DNA Methylation in Age Prediction



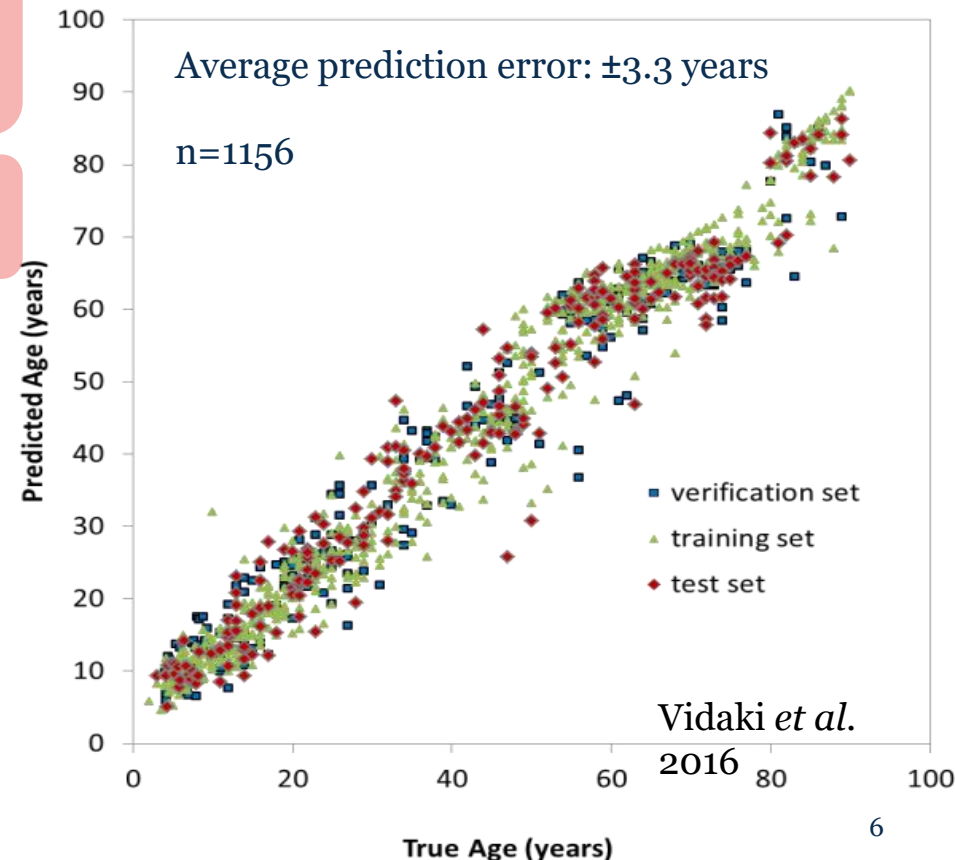
Initial Prediction Model

353 CpGs 'Epigenetic Clock' - Horvath
2013

12 age-related CpGs

Publicly available **methylation data**
deriving from the Illumina 27K and 450K
BeadChip arrays

Statistical modelling using artificial
neural networks



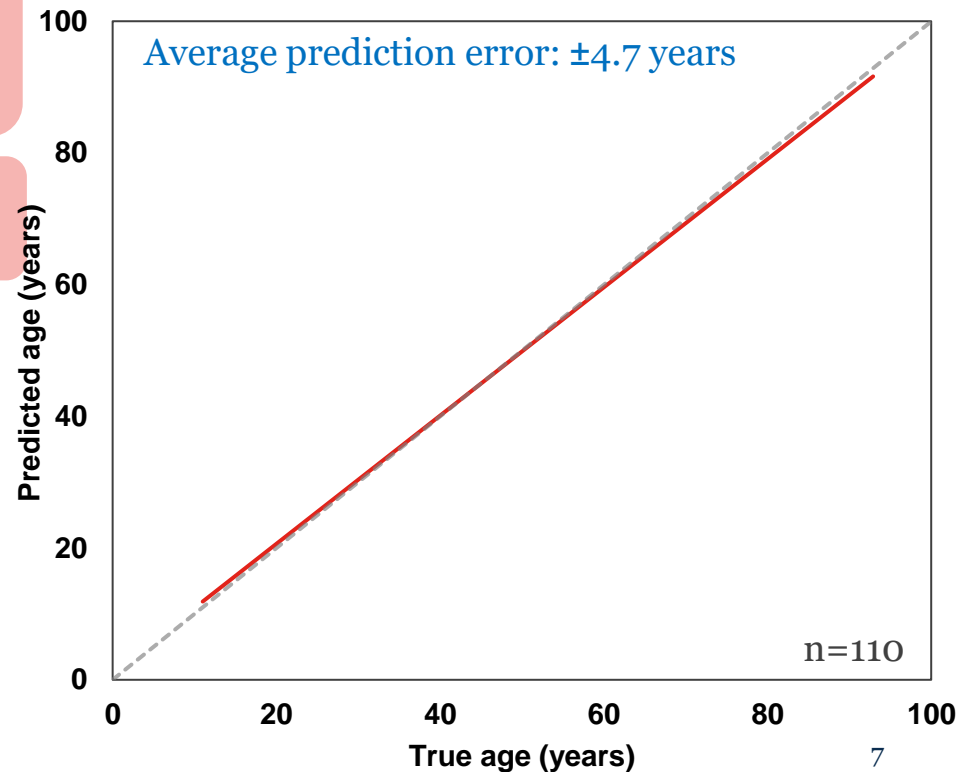
Applying the method to real samples

110 whole blood samples

Targeted DNA methylation quantification using the Illumina MiSeq technology (50ng/sample)

67 samples used for **training**, 10 for **validation** and 33 for **blind testing**

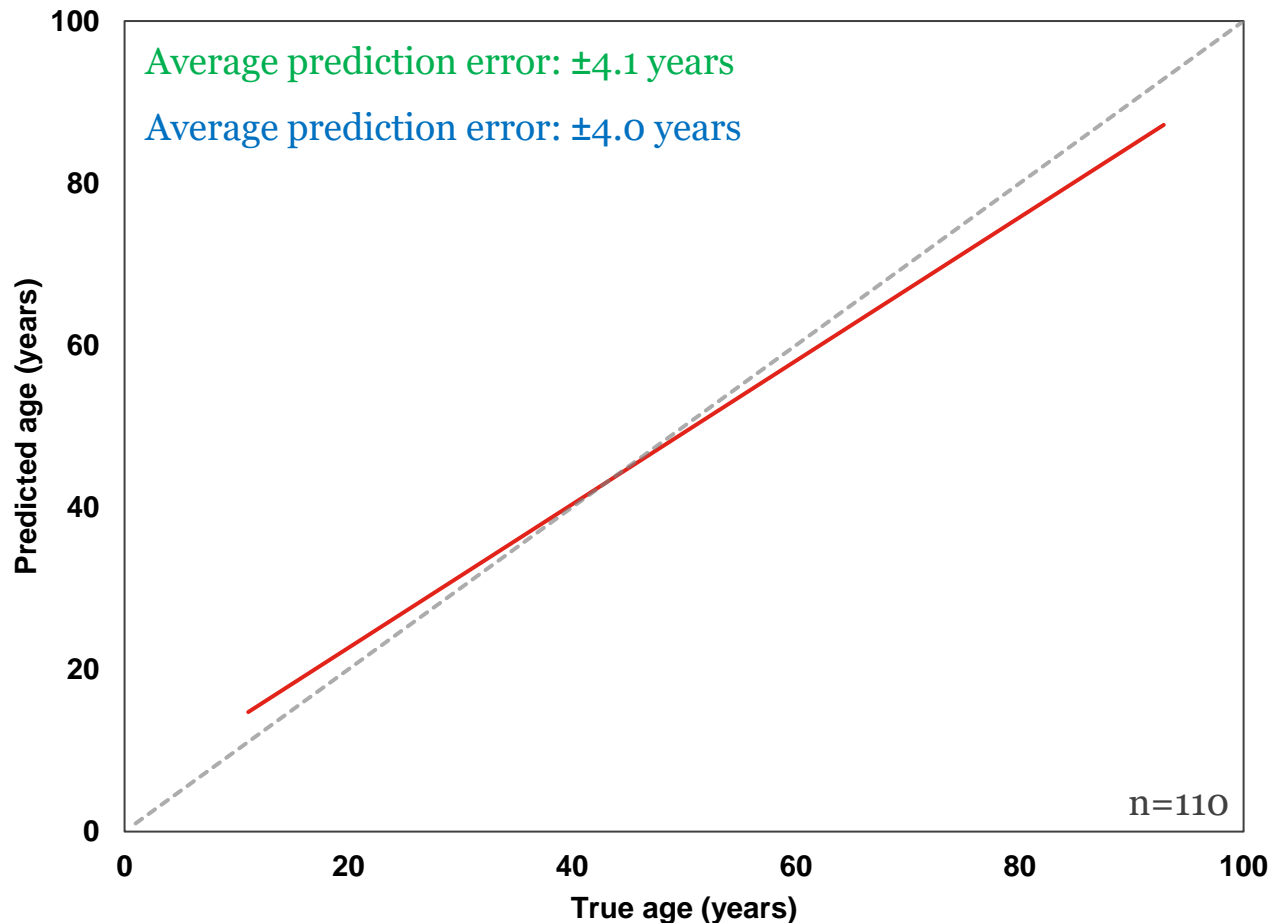
Average prediction error: ± 4.7 years



Statistical modelling

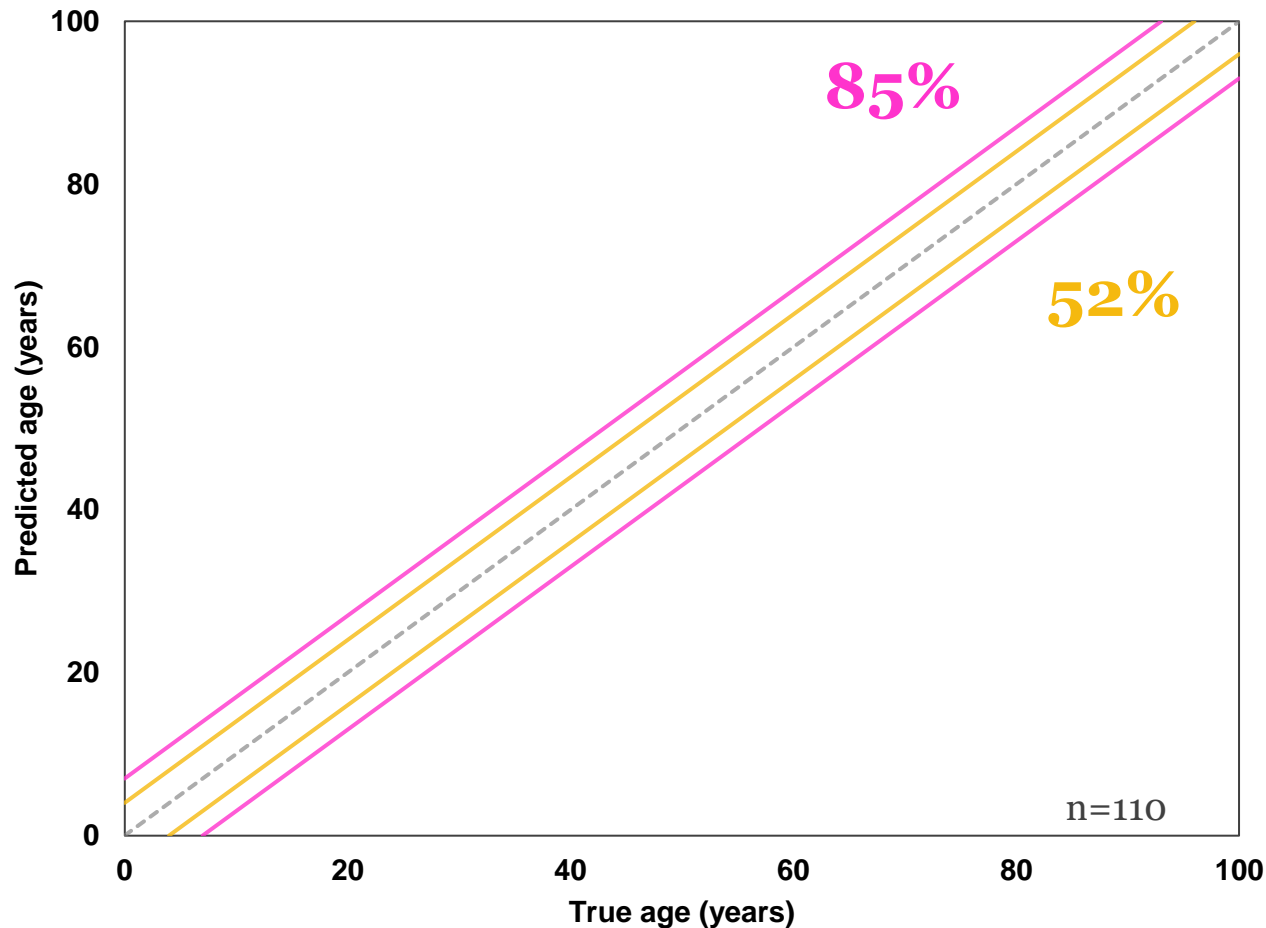
16 additional statistical approaches were tested together with the initial generalised regression model.

A support vector machine model with polynomial function was selected



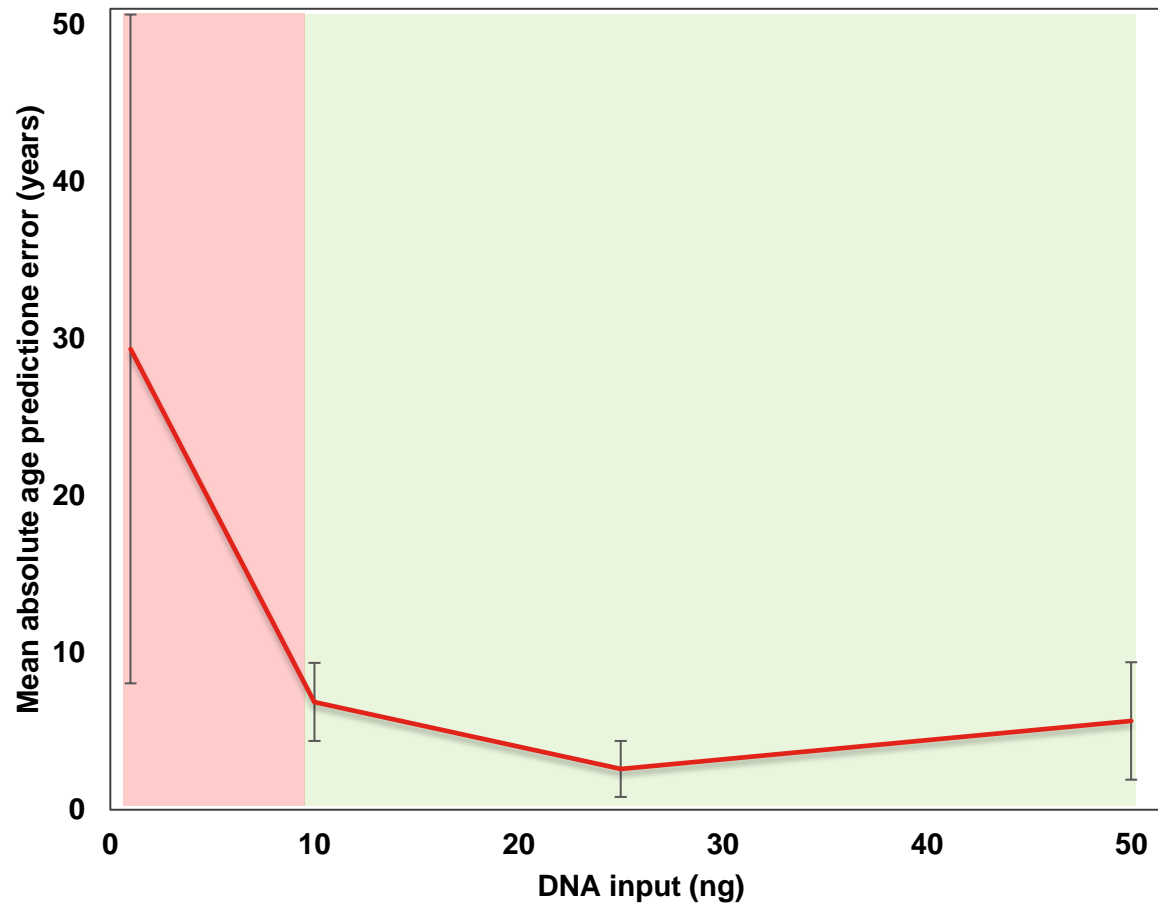
Statistical modelling

52% of the samples in the blind test predicting with **an absolute error of less than 4 years** and **85%** with an error of less than 7 years



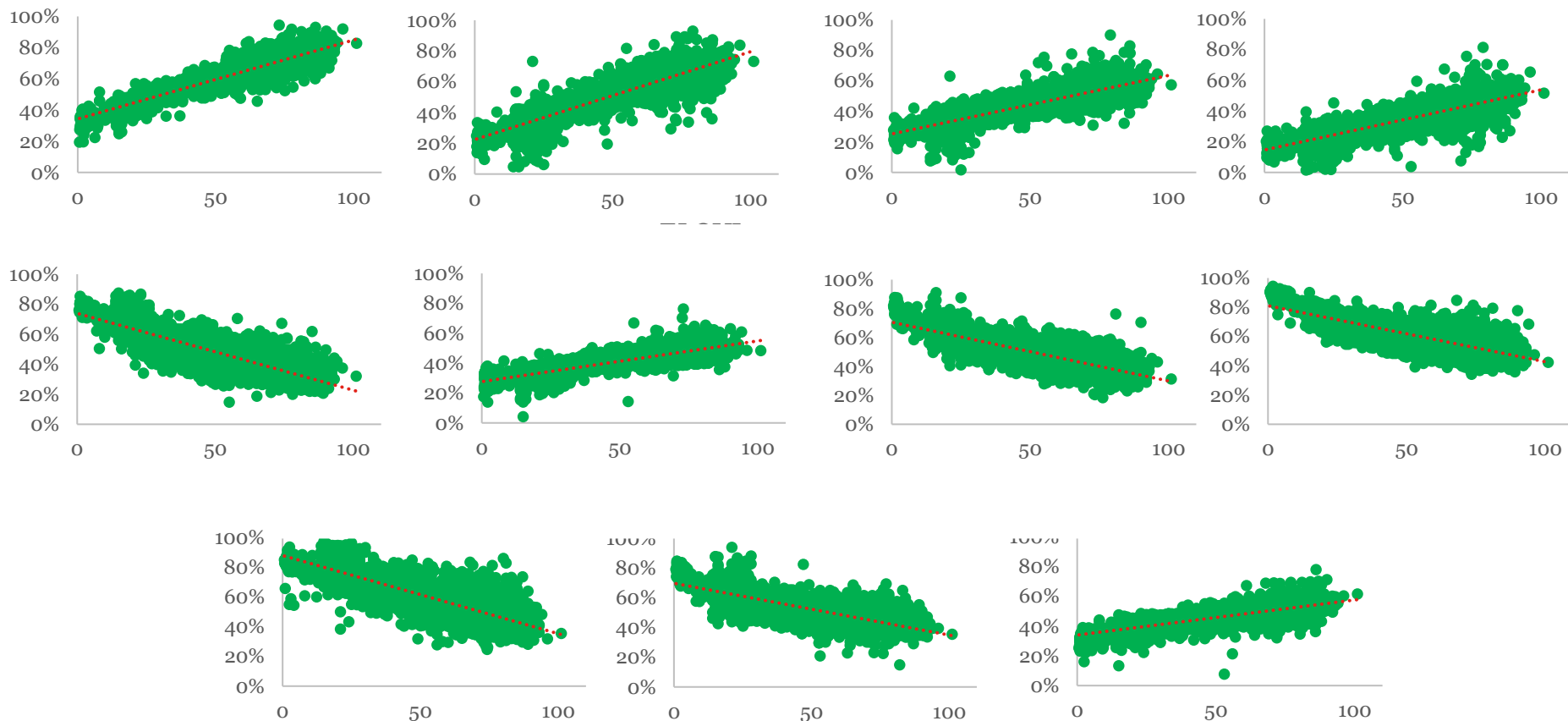
Sensitivity

The **sensitivity** of the method was also assessed using real samples. Below **10ng** the variation increases significantly.



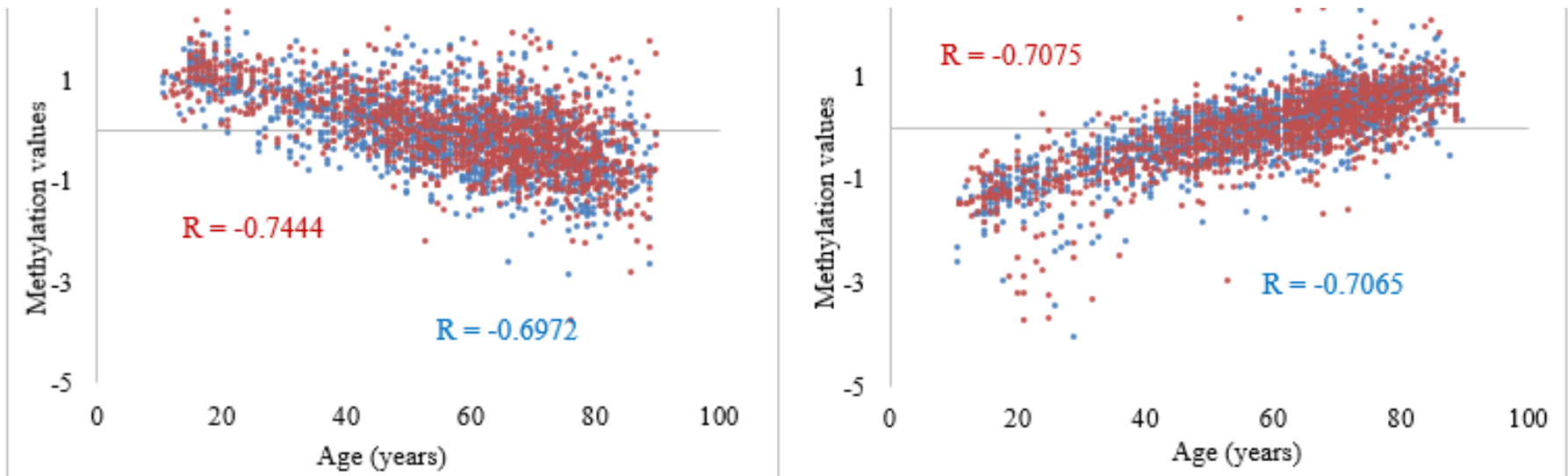
New age-related CpG sites:

from >6000 in 1 or more studies or prediction models

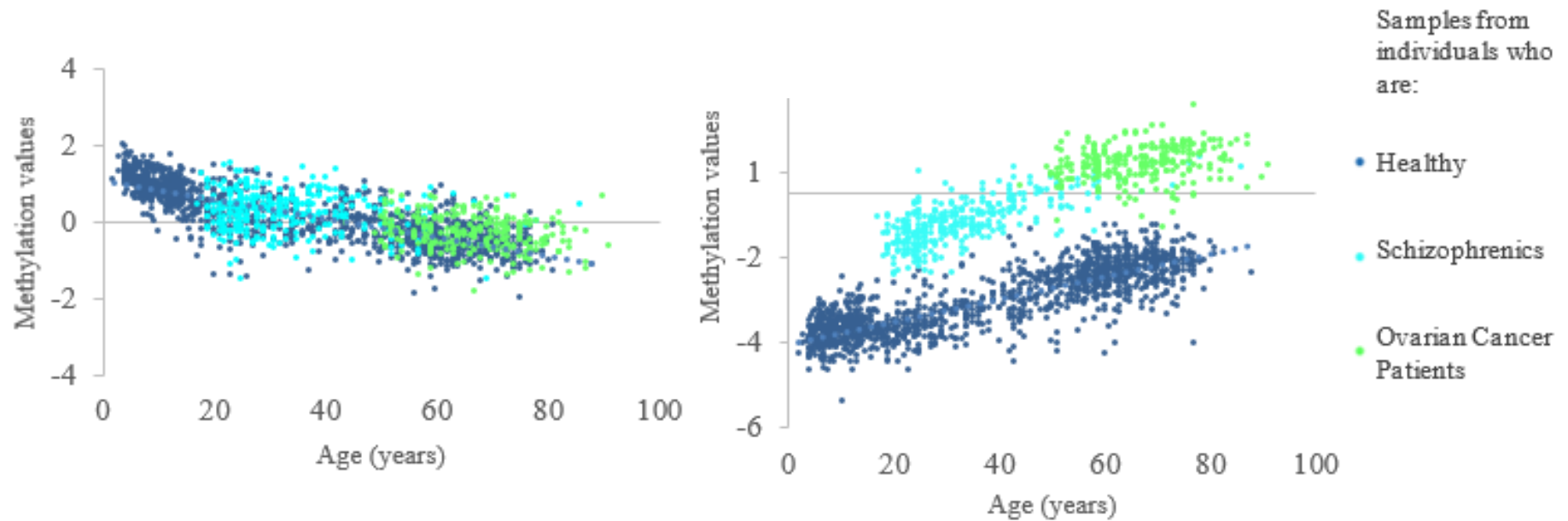


Top 11 CpGs

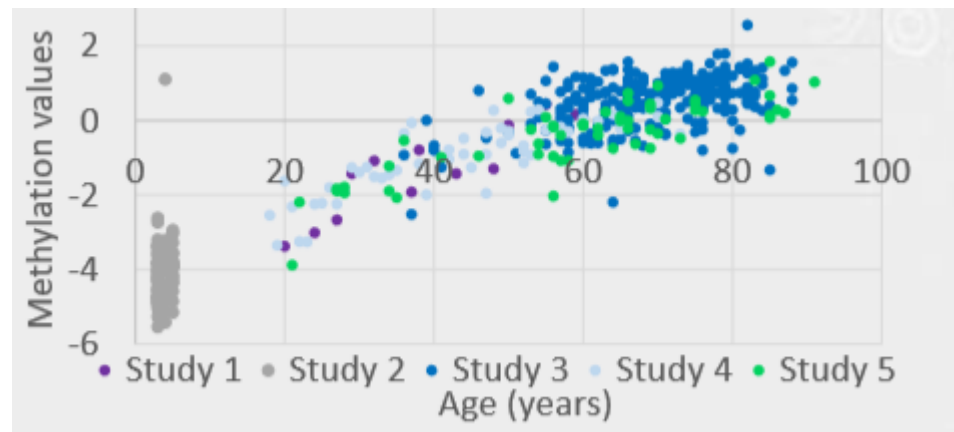
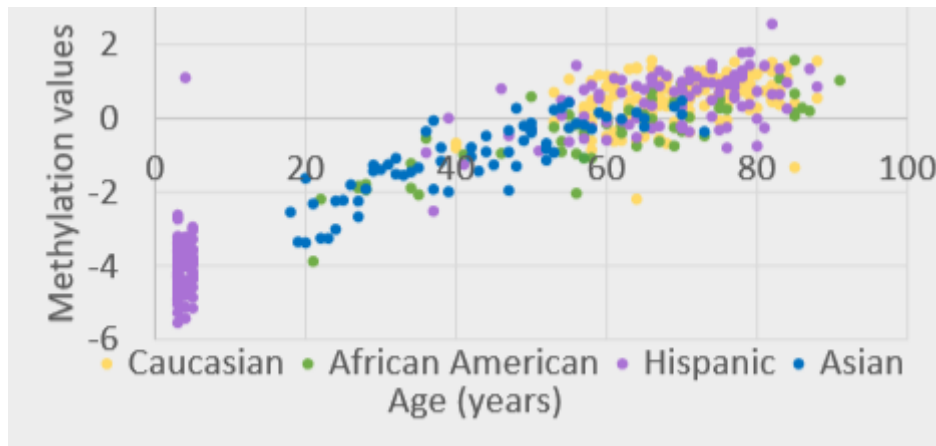
Effect of sex



Effect of disease



Effect of geographical ancestry



Skin colour – using HirisPlex-S snps

Ruth Hatton



Skin colour – making use of HirisPlex-S





Contents lists available at ScienceDirect

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Research paper

The HirisPlex-S system for eye, hair and skin colour prediction from DNA: Introduction and forensic developmental validation



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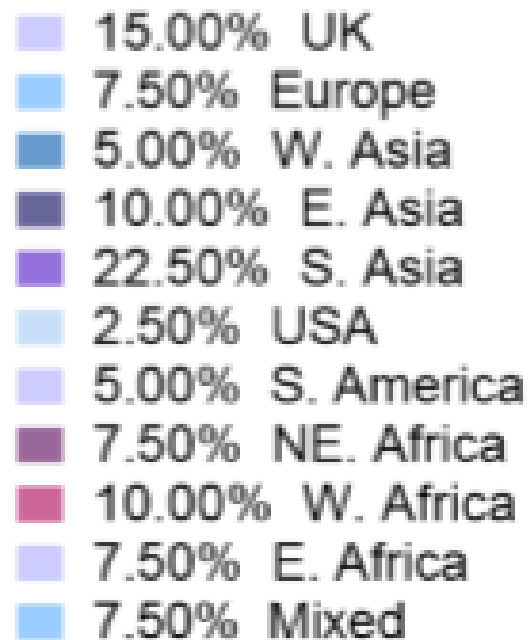
ⁱ Central Forensic Laboratory of the Police, Warsaw, Poland

HirisPlex-S Eye, Hair and Skin Colour DNA Phenotyping Webtool

<https://hirisplex.erasmusmc.nl/> ▼

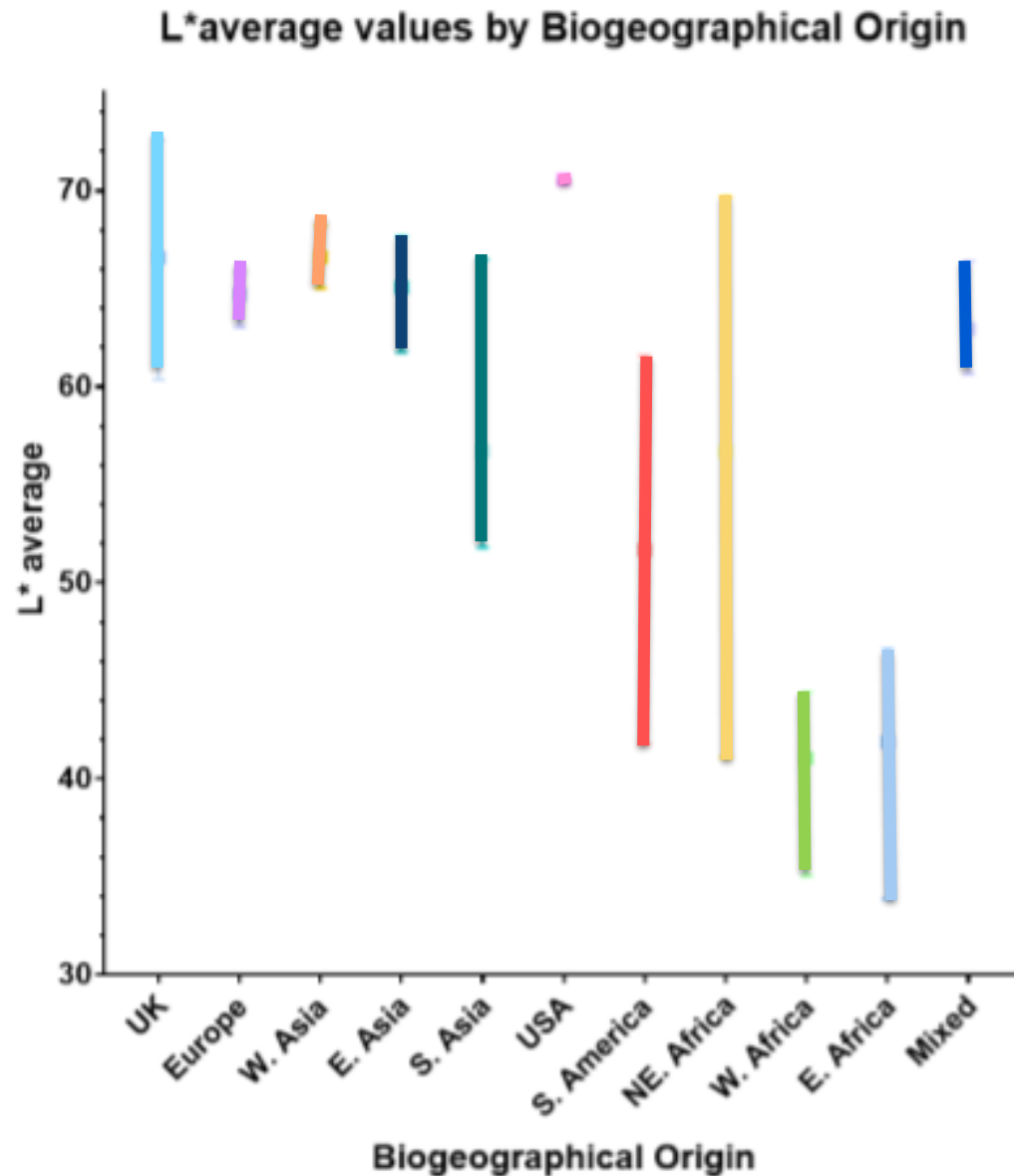
Skin study

Biogeographical Origin of Participants



Total = 40 Participants

Skin colour density



Skin colour scales

Fitzpatrick Scale	1 Always burns Never tans	2 Usually burns Minimal tan	3 Mild burn Tans uniformly	4 Minimal burn Always tans well	5 Very rarely burns Tans very easily	6 Never burns
Von Luschan Scale	0 - 6	7 - 13	14 - 20	21 - 27	28 - 34	35 - 36
HirisPlex-S Scale	Very Pale	Pale	Intermediate		Dark	Dark-Black
Participant Example						

A scatter plot showing the relationship between the Fitzpatrick score (Y-axis) and the L* average (X-axis). The Y-axis ranges from 0 to 6, and the X-axis ranges from 30 to 75. Blue dots represent individual data points. A solid red line represents the linear regression, and two dashed red lines represent the confidence intervals. The data shows a clear negative correlation, with the regression line starting at approximately (35, 6.5) and ending at (75, 1.5).

Using the web tool

A	SKIN	Probability	E	SKIN	Probability
	Very Pale	0.101		Very Pale	0
	Pale	0.117		Pale	0
	Intermediate	0.782		Intermediate	0
	Dark	0		Dark	0.001
	Dark-Black	0		Dark-Black	0.999
B	SKIN	Probability	F	SKIN	Probability
	Very Pale	0		Very Pale	0.102
	Pale	0		Pale	0.25
	Intermediate	0.02		Intermediate	0.645
	Dark	0.509		Dark	0.002
	Dark-Black	0.471		Dark-Black	0.002
C	SKIN	Probability	G	SKIN	Probability
	Very Pale	0		Very Pale	0.003
	Pale	0		Pale	0.011
	Intermediate	0		Intermediate	0.303
	Dark	0.001		Dark	0.205
	Dark-Black	0.999		Dark-Black	0.478
D	SKIN	Probability	H	SKIN	Probability
	Very Pale	0.005		Very Pale	0
	Pale	0.364		Pale	0
	Intermediate	0.607		Intermediate	0.039
	Dark	0.02		Dark	0.817
	Dark-Black	0.004		Dark-Black	0.144

Biogeographic ancestry

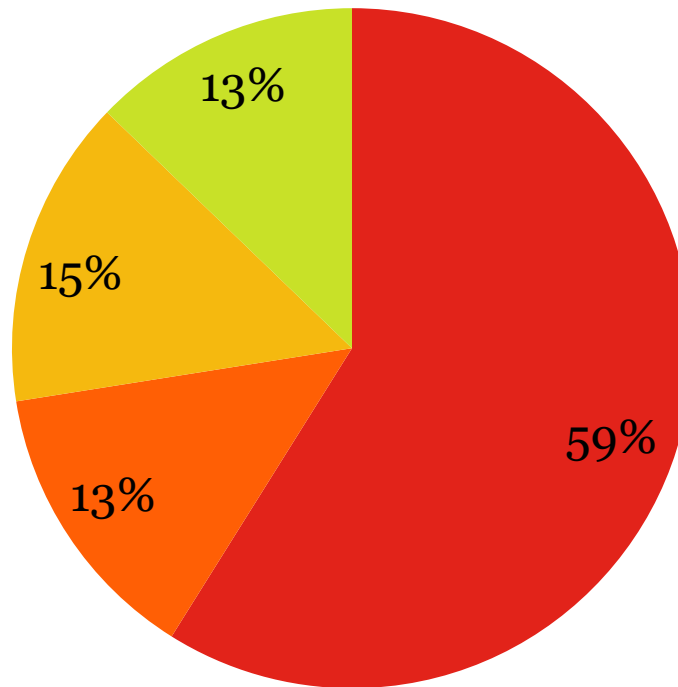
David Ballard and Laurence Devesse



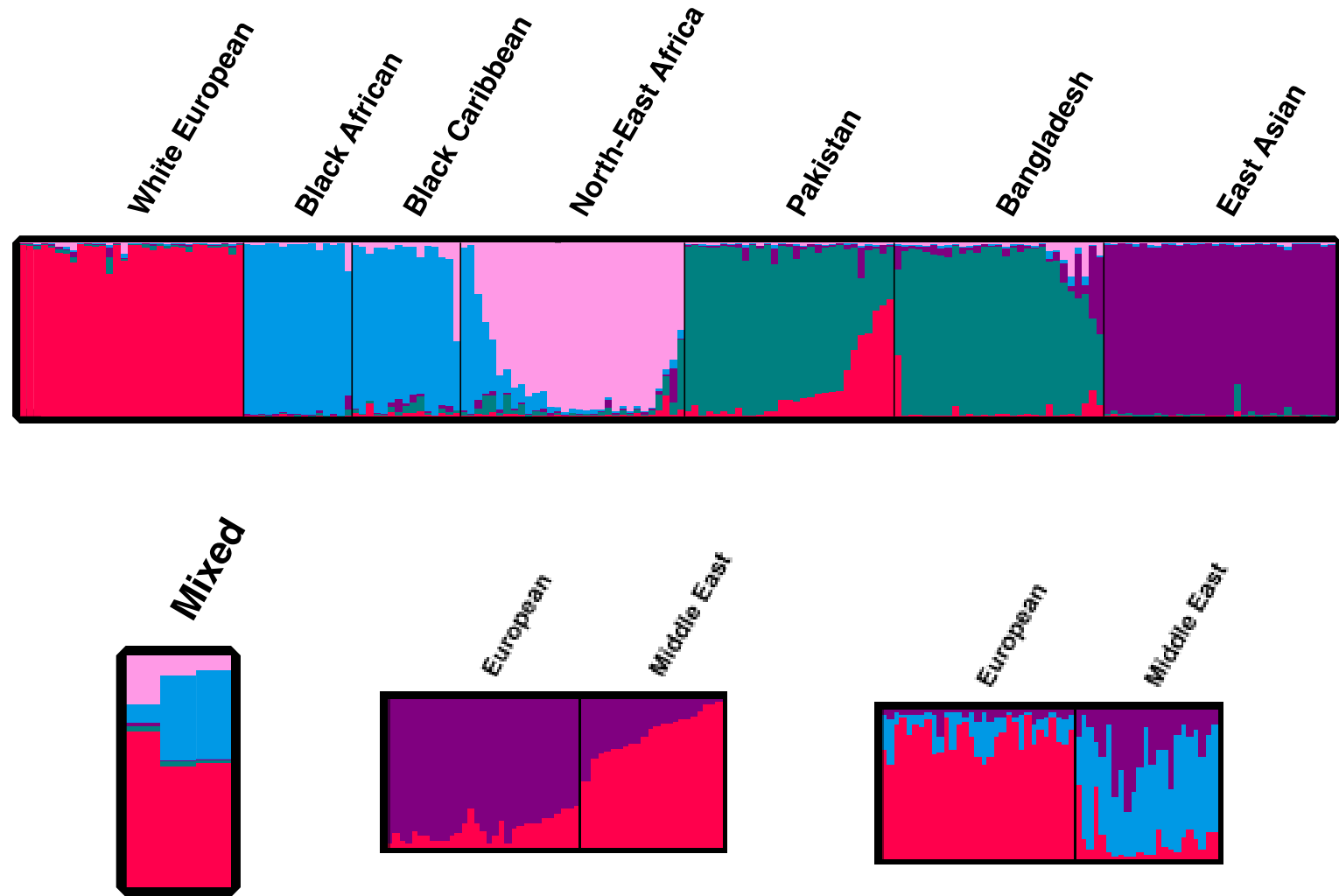
2011 census – London

Percentage of population

■ White ■ Black ■ South Asian ■ Other



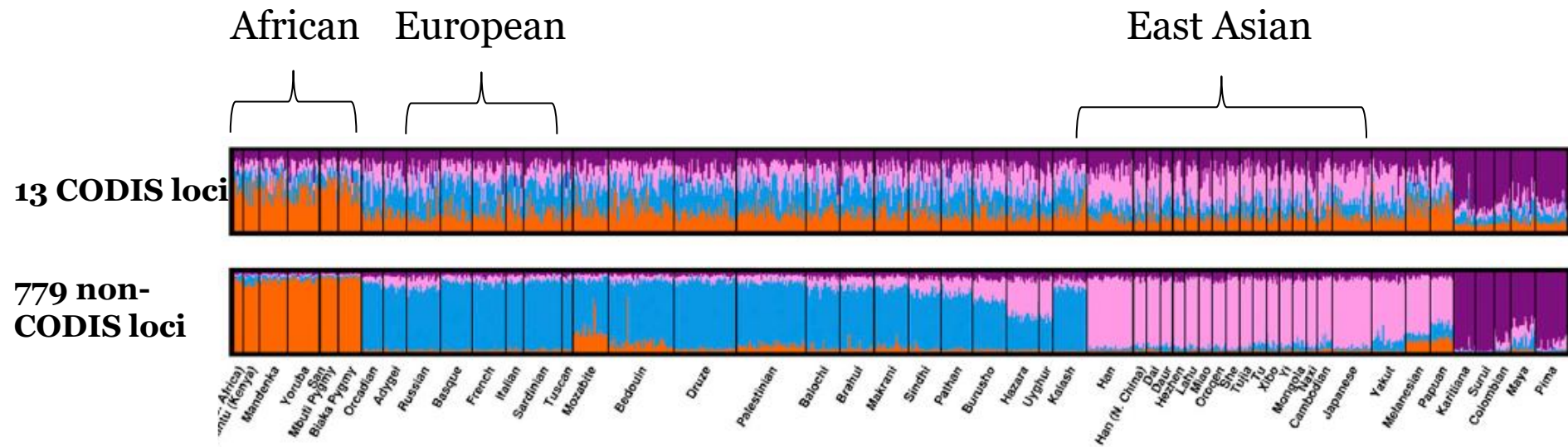
King's Ancestry SNP Panel



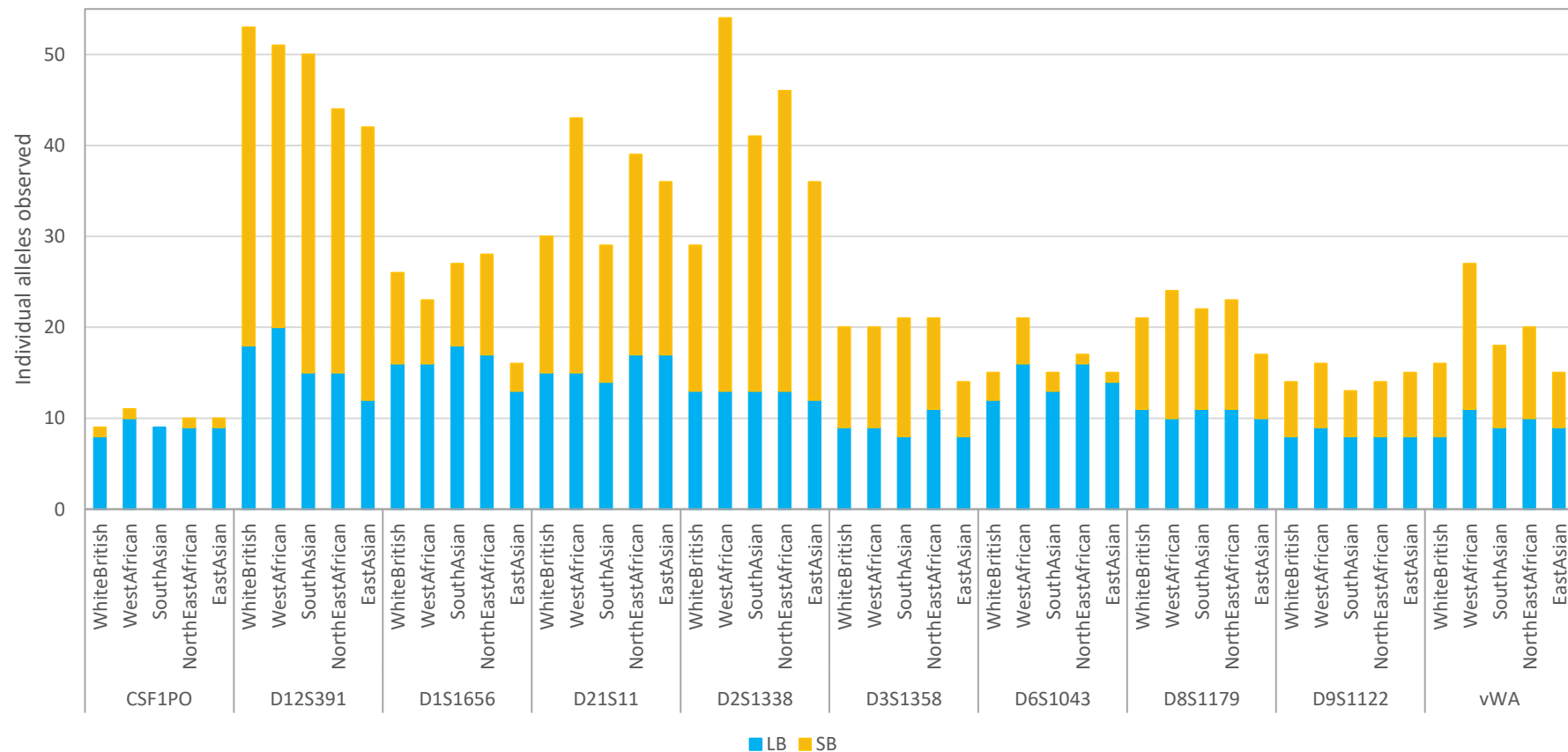
Using STRs

Using STRs for ancestry estimation

- Rosenberg (2016) showed that the standard 13 CODIS markers have ancestry information comparable to a random marker set
- Much larger marker set required to adequately differentiate continental populations

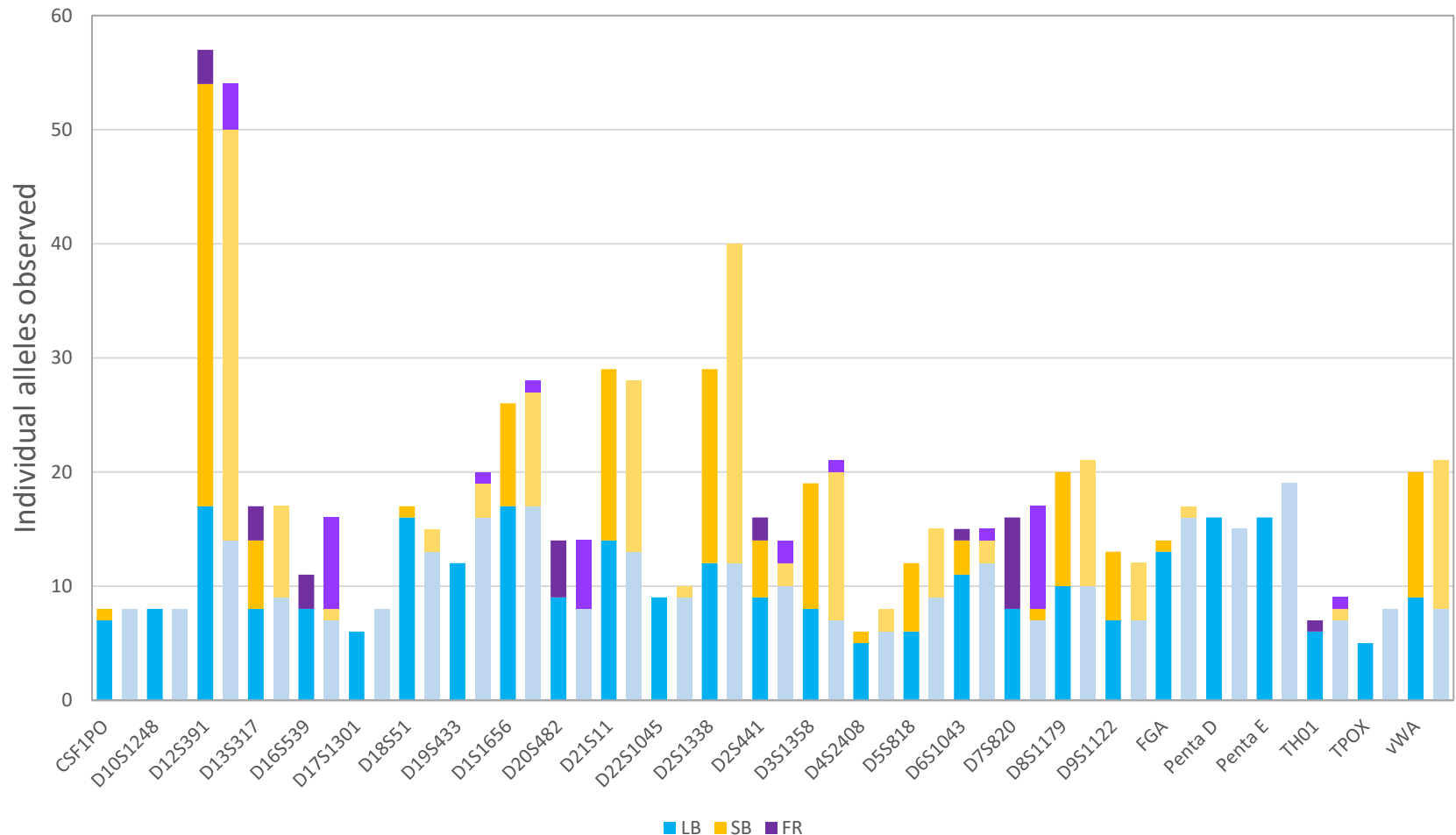


Sequence variation

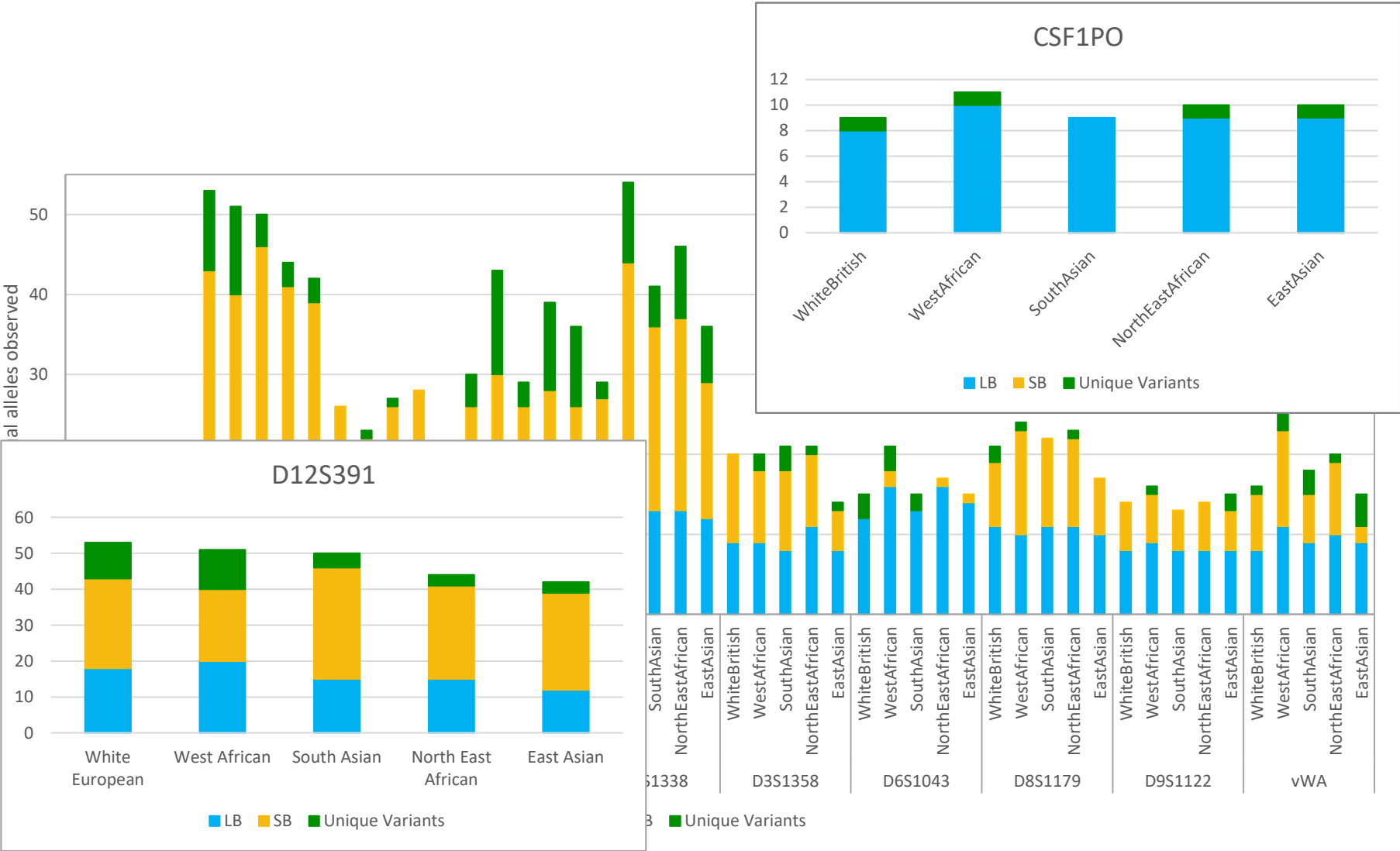


Flanking region variation

- Variation has also been observed in the flanking region of STRs
- Data from two population groups: White European and South Asian

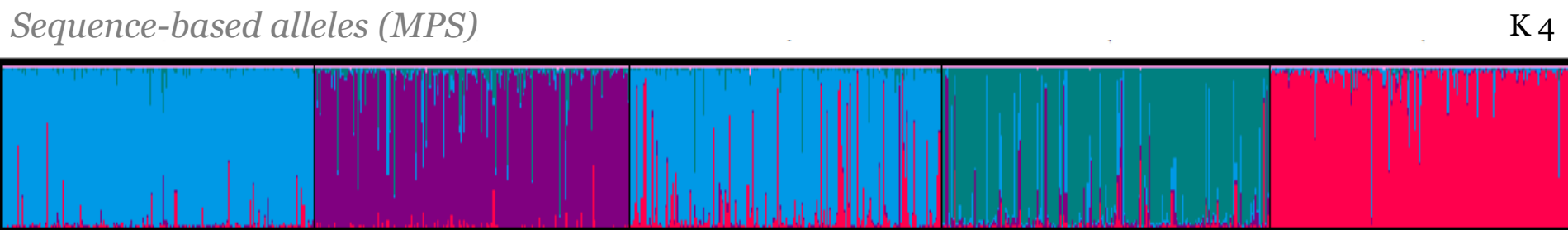
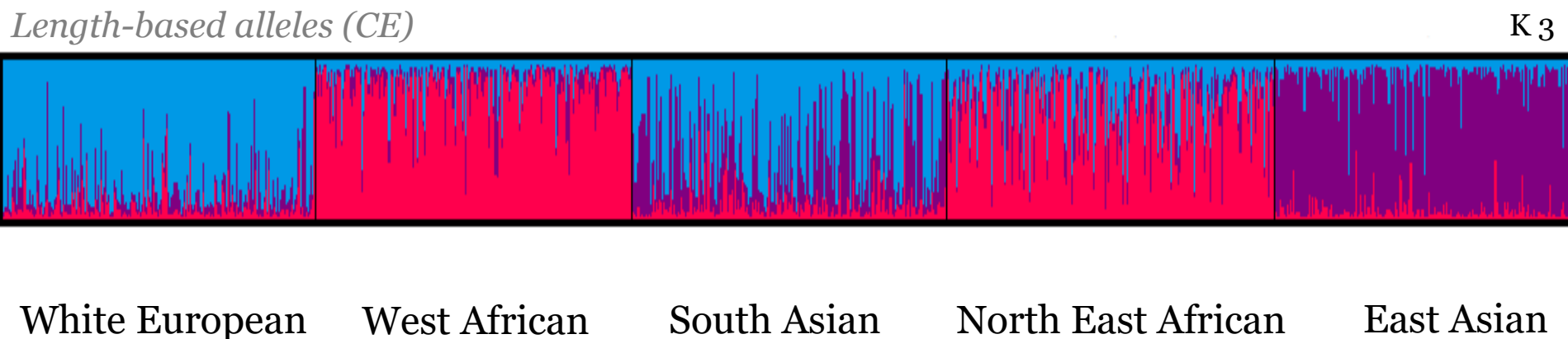


Population specific variation



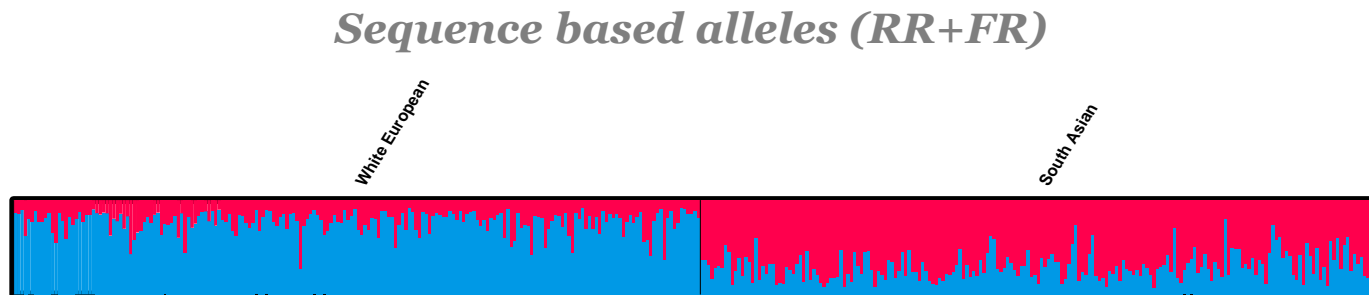
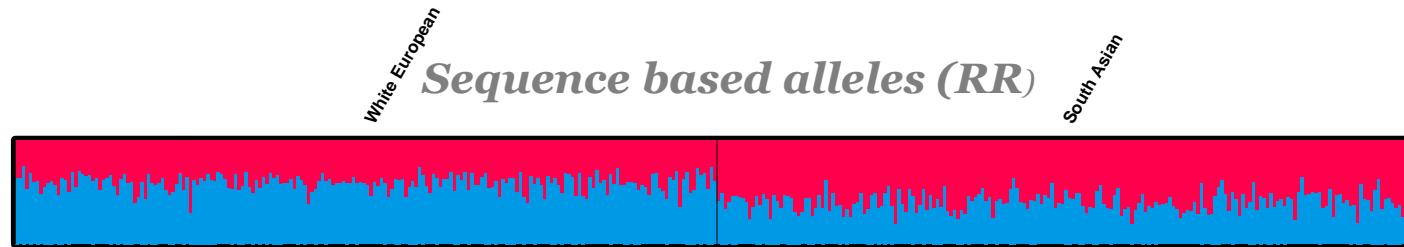
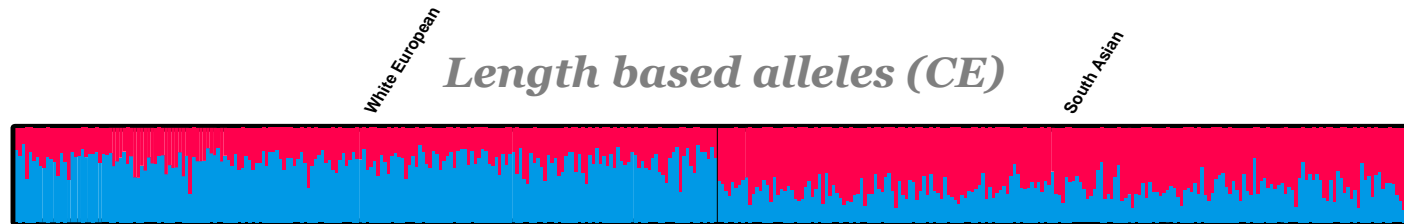
Bio-geographical ancestry inference

- Can we use sequence variation to infer ancestry?



- Population differentiation using sequence-specific alleles from 27 **forensic autosomal STRs**

Using the flanking region sequences



How should we report this information?



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Research Article

Implementing a biogeographic ancestry inference service for forensic casework

The Centre of Forensic Sciences has validated the Precision ID Ancestry Panel on the Ion S5TM Massively Parallel Sequencing instrument for use in forensic casework. The focus of this paper is the development of reporting guidelines for implementation of the biogeographic ancestry inference service based on the Admixture Prediction results produced using the Torrent SuiteTM Software (Thermo Fisher Scientific). The Admixture Prediction algorithm estimates the genetic ancestry of a sample using seven root populations (Europe, East Asia, Oceania, America, Africa, South Asia, and Southwest Asia). For individuals that declared a single ancestry, there was a high correlation between the declared ancestry and the ancestry predicted by the algorithm. However, some individuals with declared ancestries of Southern Europe, Southwest Asia, South Asia and Horn of Africa had Admixture Predictions that were composed of two or more root populations at 20% or greater. For individuals with known admixed ancestry, the major component of their declaration was included in their results in all but one case. Based on these results, reporting guidelines were developed and subsequently evaluated using the Admixture Predictions of additional samples. This paper discusses the development and evaluation of these reporting guidelines, along with an implementation plan for forensic casework.

Keywords:

Ancestry informative SNP / Biogeographic ancestry / Ion S5 / Massively Parallel Sequencing / Precision ID Ancestry Panel

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Additional supporting information may be found online in the Supporting Information section at the end of the article.

Admixture predictions based on root populations

Root populations:

Europe, Oceania, East Asia, Africa, South Asian, America, Southwest Asia

Declared ancestry (no admixture)	Percentage concordance with prediction
Northern Europe	80 – 100
Southern Europe	55-95
Sub-Saharan Africa	70-100
North-East Africa	30-70
North Africa	8-10
East Asia	80-100
South Asia	35-95
Middle East	30-90

Reporting guidelines (CFS Ontario)

- **Single inclusion**
 - One root population >80%
 - All other populations 15% or less
- **Single mixed inclusion (single root or mixed with at least one other)**
 - One root 55% - 75%
 - All other populations 15% or less
- **Double inclusion (single root of either, or mixture of both)**
 - Two roots 20% or greater
- **Uninformative**
 - At least three roots 20% or greater

Predictions – King’s Forensics – using LR

		Prediction -log(likelihood)					
		Caucasian	West African	Somalian	Chinese	Pakistan	Bangladesh
1	Caucasian	20.04	76.27	64.14	76.89	29.25	41.49
2	Caucasian	26.46	82.58	70.95	80.87	36.47	46.33
3	Caucasian	24.78	78.95	50.31	68.15	28.19	35.89
4	Caucasian	21.59	77.62	62.11	74.26	30.27	41.65
5	Caucasian	24.51	79.22	63.23	68.49	33.80	44.80
6	Caucasian	23.59	70.57	49.79	61.91	31.40	36.78
7	Caucasian	23.23	72.29	54.43	76.80	34.19	43.51
8	Caucasian	31.21	71.97	53.91	63.99	36.74	42.77
9	Caucasian	20.05	96.73	67.71	75.37	33.46	46.04
10	Caucasian	21.97	83.32	59.99	71.56	30.91	39.58
11	Caucasian	23.05	76.33	56.06	64.81	27.20	37.74
12	Caucasian	20.34	77.33	63.08	79.41	33.16	42.50
13	Caucasian	28.89	72.35	58.27	61.81	33.54	36.48
14	Caucasian	29.49	79.64	70.68	87.78	36.81	50.01
15	Caucasian	35.29	73.78	59.55	75.00	38.53	48.93
16	Caucasian	19.09	78.45	62.98	75.73	27.90	38.99
17	Caucasian	17.25	83.56	72.10	86.35	32.67	47.01

Predictions

		Prediction -log(likelihood)					
		Caucasian	West African	Somalian	Chinese	Pakistan	Bangladesh
21	West_Africa	129.59	29.08	39.05	71.59	84.32	67.87
22	West_Africa	103.78	24.57	28.61	70.72	70.82	59.88
23	West_Africa	105.33	22.64	34.00	77.97	75.89	62.81
24	West_Africa	123.72	23.96	35.07	82.72	89.09	76.34
25	West_Africa	99.70	21.27	32.26	65.46	71.97	59.74
26	West_Africa	91.74	26.65	35.08	72.01	70.68	59.43
27	West_Africa	112.01	25.02	37.69	73.69	82.36	67.61
28	West_Africa	108.80	28.38	35.62	68.26	78.25	65.91
29	West_Africa	114.97	24.47	36.72	76.64	84.48	69.92
30	Somalian	81.39	38.27	29.87	61.67	59.47	50.69
31	Somalian	94.85	43.40	26.56	54.77	62.44	54.84
32	Somalian	108.20	42.23	31.40	59.19	65.17	52.78
33	Somalian	88.27	29.63	32.55	73.85	72.00	61.84
34	Somalian	70.87	36.72	27.32	64.06	52.90	48.06
35	Somalian	89.28	46.44	34.87	61.69	57.38	48.81
36	Somalian	67.13	52.22	27.47	48.78	46.13	39.76
37	Somalian	77.53	49.74	32.59	57.76	49.62	44.32
38	Somalian	85.90	49.60	31.39	59.73	59.35	53.82

Predictions

		Prediction -log(likelihood)					
		Caucasian	West African	Somalian	Chinese	Pakistan	Bangladesh
39	Chinese	101.79	88.08	73.23	20.24	48.74	37.64
40	Chinese	89.39	91.11	71.08	29.18	48.10	41.26
41	Chinese	96.65	96.60	79.97	24.22	49.29	38.24
42	Chinese	94.42	88.14	68.97	22.75	44.35	35.05
43	Chinese	83.64	76.89	69.38	23.64	46.19	36.46
44	Chinese	96.04	91.07	76.82	18.77	43.00	36.28
45	Chinese	99.57	73.02	61.39	24.55	53.47	38.91
46	Pakistan	47.71	75.54	55.32	48.34	24.04	26.44
47	Pakistan	58.00	90.15	66.12	62.48	28.42	31.37
48	Pakistan	63.56	68.57	54.07	39.78	34.56	29.83
49	Pakistan	33.58	70.42	47.25	56.49	23.67	26.54
50	Bangladesh	60.23	62.97	50.51	43.94	32.69	22.12
51	Bangladesh	56.14	80.31	45.06	44.10	32.72	23.58
52	Bangladesh	51.34	56.49	39.65	42.62	34.20	30.67
53	Bangladesh	56.92	65.92	45.83	39.75	30.02	22.99

How good is the prediction? – blind test

Prediction Given	Caucasian	African West	African N/NE	African region unclear	South Asian	East Asian	Unclear
Caucasian	84%						15%
Nigerian/Jamaican		50%		47%			3%
Somali			71%	23%			6%
Chinese						94%	6%
Pakistan					100%		0%
Bangladesh					90%		10%

What is the unclear?

Mixed population within parents/grandparents

Close population group not represented in model

STRUCTURE, Y STRs, MtDNA

Conclusions



Recommendations for phenotypic applications

- Methodologies should be validated using samples representative of the target population groups
- Support validation with blind tests
- Purpose of test must be to aid an investigation in order to assist prioritisation, and only when:
 - An STR profile, suitable for to load to a database, has revealed no hits
 - Only if sufficient of a high quality DNA extract is available

Recommendations for guidelines

- Convey the limitations of the test and mitigate the risks of misinterpretation through:
 - A statement explaining the limitations of the relevant tests to form part of the service request
 - Where possible the risk of a wrong attribution should be stated (blind tests can assist here)
 - Make it clear to the investigator that no person should be excluded from further investigation as a consequence of the test result
 - Provide a glossary to explain and clarify;:
 - Differences between geographic ancestry and where a person was born
 - Genetic ancestry and cultural ancestry (ethnicity)
 - Expected variability within the population
 - Model will depend on accuracy of declarations made by test subjects
 - Caution to be considered in admixed ancestry individuals

Thank you

