

# GENERATING INVESTIGATIVE LEADS FROM FORENSIC DNA DATA:

## Mapping Y-STR Profiles to Ancestral Haplogroups



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### FROM STATIC DNA EVIDENCE TO FORENSIC INTELLIGENCE

While Y-chromosome short tandem repeat (Y-STR) typing is the routine forensic analysis for individual identification, it could not provide investigative leads without a direct match from a previously populated database. This research bridges that gap by using machine learning to predict Y-chromosome single nucleotide polymorphism (Y-SNP) haplogroup, transforming what was static DNA evidence into generative intelligence for paternal ancestry and cold case leads.

### PROJECT OBJECTIVES

- To predict Y-SNP haplogroups directly from forensic Y-STR profiles using machine learning.
- To demonstrate the use of SHAP (SHapley Additive exPlanations) for model interpretability and judicial transparency.

We engineered a machine learning pipeline using **stratified sampling** and **class weighting** to address extreme population imbalances and ensure high precision for minority lineages.

### THE DATASET AND DATA PROCESSING PIPELINE

**4,064**

unique Y-STR profiles characterized by

**27 Y-STR**

genetic markers to associate with

**13 Y-SNP**

**Haplogroups**

representing paternal lineages

sourced from the comparative study by Song et al. (2024) YHP: Y-chromosome Haplogroup Predictor for predicting male lineages based on Y-STRs.

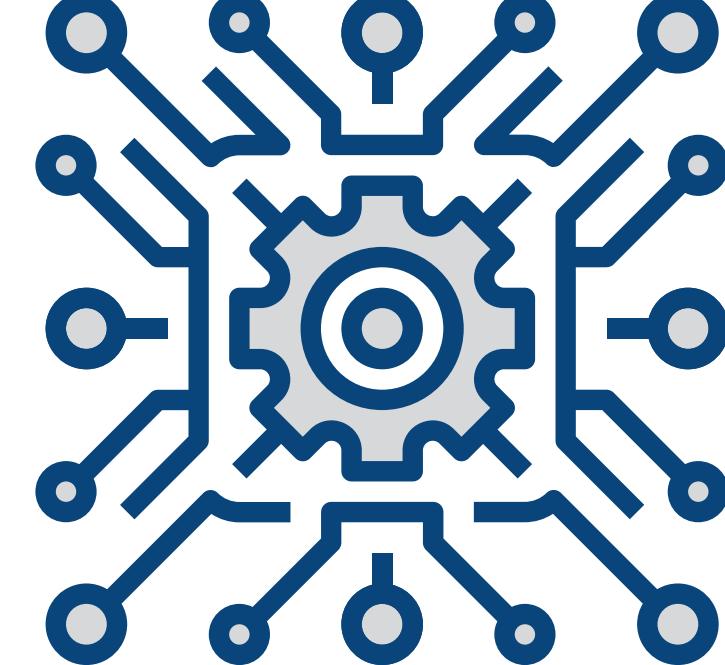
#### DATASET PRE-PROCESSING

**INPUT:** Forensic Y-STR Data with known Y-SNP Haplogroups Training (75%) and Hold-out Dataset (25%).



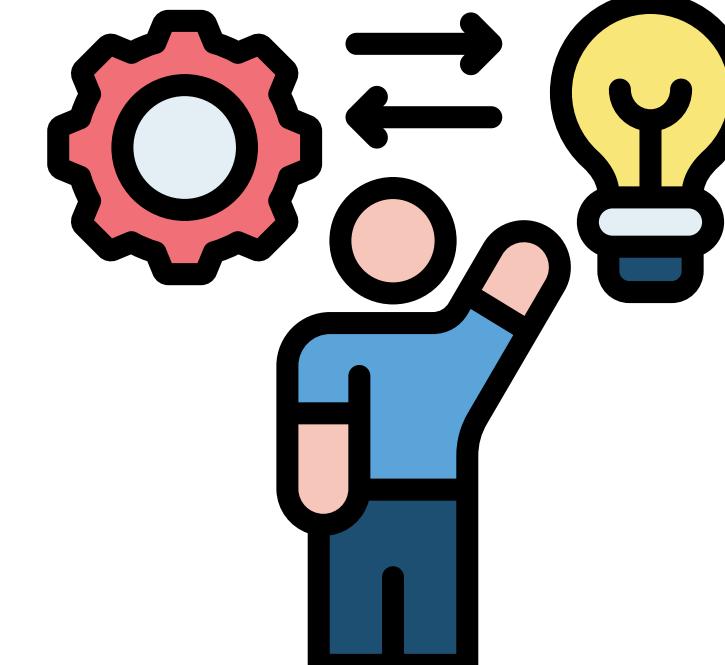
#### MACHINE LEARNING MODEL SELECTION

**PROCESS:** Stratified sampling used to survey best model among 8 supervised classifiers.



#### MODEL INTERPRETABILITY

**ANALYSIS:** Use SHAP Values to identify key Y-STR markers crucial in predicting the corresponding haplogroup.



#### INVESTIGATIVE LEAD GENERATION

**OUTPUT:** Probabilistic reporting of Y-SNP Haplogroup from Y-STR input for actionable lead.

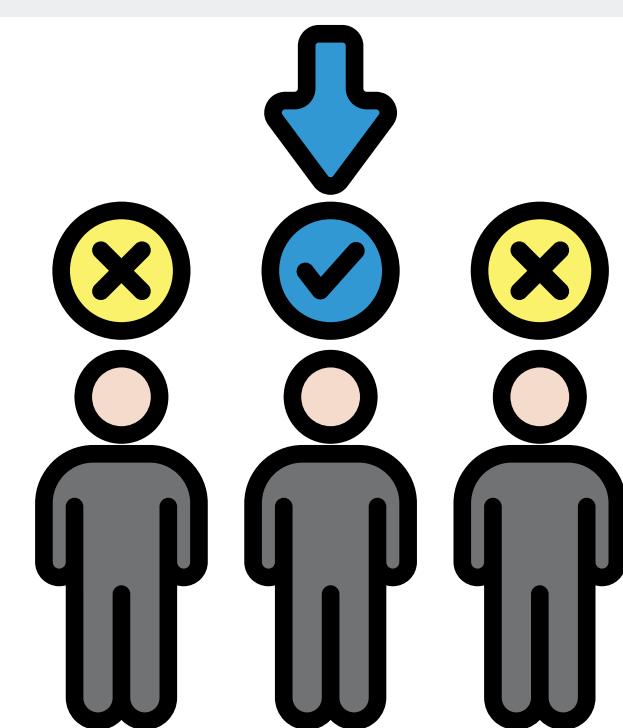
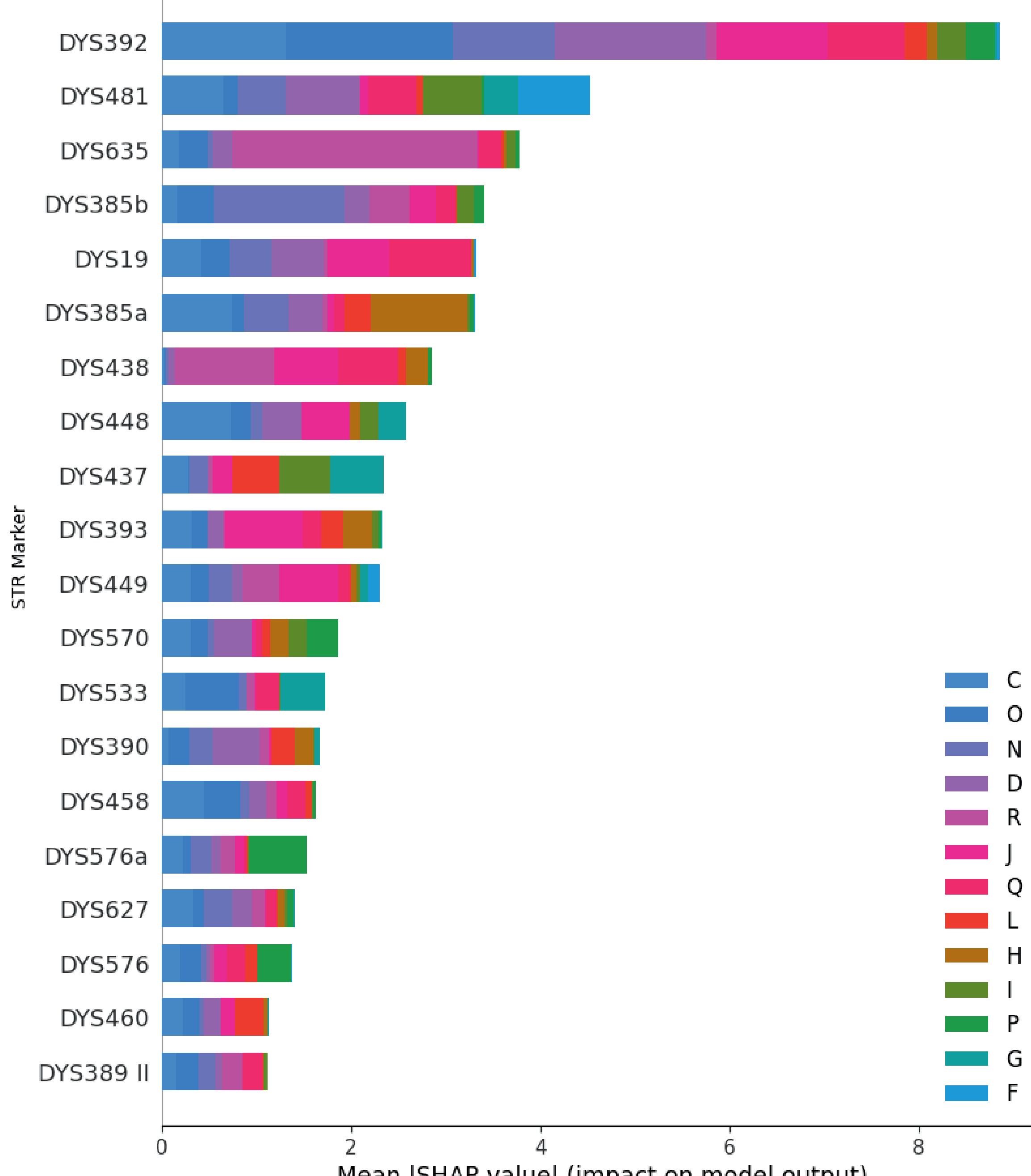


Figure 1. Design of data processing pipeline used in the study.

### KEY FINDINGS: MODEL PERFORMANCE & INTERPRETABILITY FOR LAW ENFORCEMENT APPLICATIONS

Machine Learning	Model Accuracy
XGBoost	0.9698
Random Forest	0.9679
SVM	0.9623
kNN	0.9616
LDA	0.9438
Gaussian NB	0.9351
Decision Tree	0.9342
Elastic Net	0.2029

#### SHAP FEATURE IMPORTANCE BY CLASS



#### STEPPING INTO THE FUTURE OF FORENSIC INVESTIGATIONS

- This research transforms static DNA data into actionable leads by mapping Y-STR profiles to Y-SNP haplogroups with high fidelity, achieving an overall F1-score of **0.9810**.
- By integrating **XGBoost** with **SHAP interpretability** and class weighting, the framework ensures the judicial transparency and demographic fairness essential for "Trustworthy AI" in forensic settings.
- To operationalize this tool, we recommend implementing **regional data retraining** and **independent peer-validation** of explainability results to establish a **standardized, ethical global protocol** for generative investigative intelligence.

Table 1. Machine Learning Models Performance.

The optimized XGBoost framework achieved a superior **96.98%** Accuracy and Macro F1-score of **0.9810**, drastically outperforming linear models like Elastic Net (20.29%).

Haplogroup Class	Precision	Recall	F1-Score
Major (e.g. O)	0.99	0.98	0.99
Minor (e.g. L)	0.97	0.96	0.97
Overall Model	<b>98.24%</b>	<b>96.98%</b>	<b>0.981</b>

Table 2. Summary of XGBoost Model Performance.

The model maintains high fidelity across imbalanced populations, ensuring reliable investigative leads regardless of ancestral origin.

Figure 2. Summary of SHAP Values per Y-STR genetic marker.