PREDICTING FETAL HEALTH WITH MACHINE LEARNING: HEART RATE VARIABILITY AS A NON-INVASIVE MARKER OF AUTONOMIC NERVOUS SYSTEM MATURATION

By Shane Bollinger 8/8/2025

PROJECT OVERVIEW

- Used a fetal health dataset to train ML models on CTG features.
- Gradient Boosting model: **95%** training, **90%** test accuracy.
- Interpreted key features like **HRV** as markers of **ANS development**.
- Bridges **machine learning** and **neurophysiology** for clinical and research use.

Dataset Columns

- Accelerations: Number of accelerations per second
- Fetal_movement: Number of fetal movements per second
- Uterine contractions: Number of uterine contractions per second
- Severe_decelerations: Number of severe decelerations per second
- Prolonged_decelerations: Number of prolonged decelerations per second
- Abnormal_short_term_variability: Percentage of time with abnormal short-term variability.
- Mean_value_of_short_term_variability
- Percentage_of_time_with_abnormal_long_term_variability
- Mean_value_of_long_term_variability
- Histogram_width: Range between the highest and lowest histogram values
- Histogram_min: Lowest value in the histogram
- Histogram_max: Highest value in the histogram
- Histogram_number_of_peaks: Count of distinct peaks in the histogram
- Histogram_number_of_zeroes: Number of zero-frequency bins in histogram
- Histogram_mode: Most frequent value in the histogram
- Histogram_mean: Average value of histogram data
- Histogram_median: Middle value of histogram data
- Histogram_median *(duplicate)*: Same as above
- Histogram_variance: Spread of values in the histogram
  Histogram_tendency: Skew or direction of histogram distribution
- Fetal_health: 1 - Normal; 2 - Suspect; 3 - Pathological

RESEARCH QUESTIONS

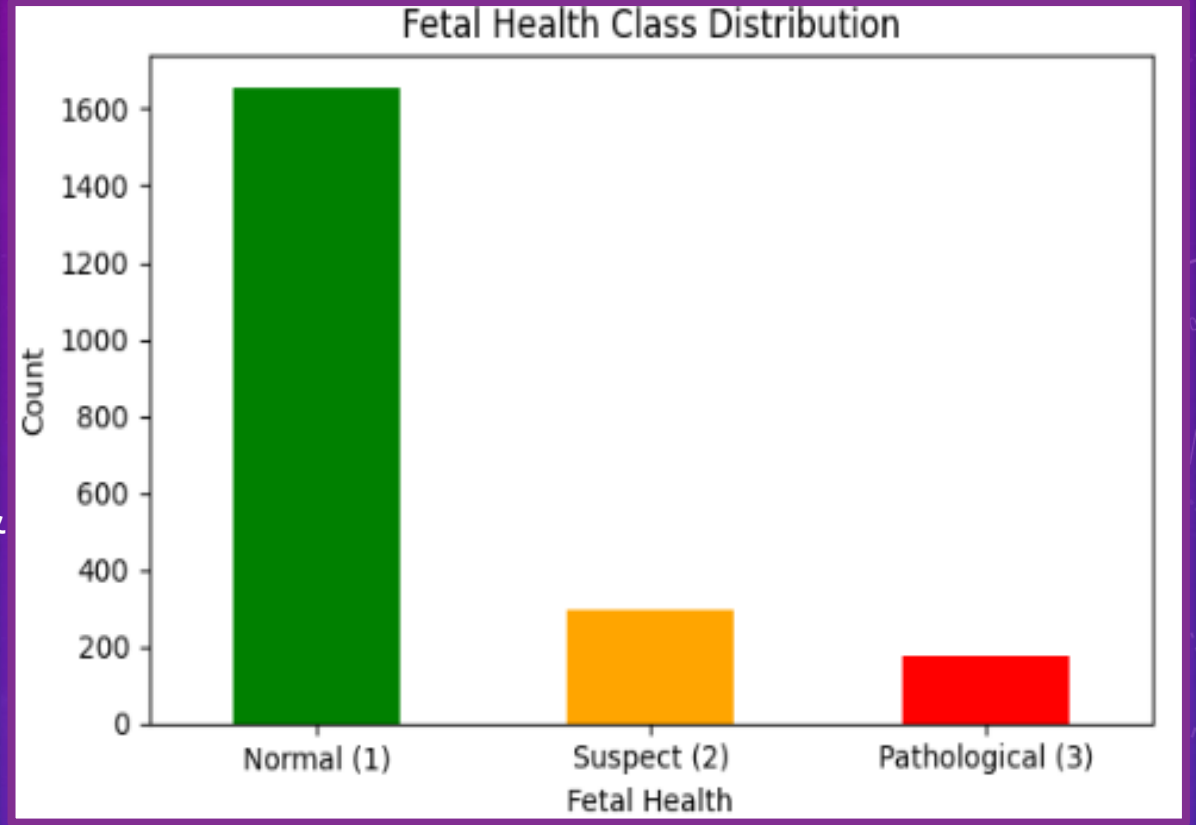
Research Question #1: Can we accurately predict fetal health status using cardiotocographic features?

Research Question #2: Can variability measures in fetal heart rate be non-invasive indicators of autonomic nervous system (ANS) maturation?

EXPLORATORY DATA ANALYSIS

Top 5 Most Correlated Features (Excluding Histograms)

- Mean_value_of_short_term_variability & light decelerations **56%**
- Fetal_health & prolongued_decelerations **48%**
- Abnormal_short_term_variability & fetal_health **47%**
- percentage_of_time_with_abnormal_long_term_variability & mean_value_of_short_term_variability **47%**
- percentage_of_time_with_abnormal_long_term_variability abnormal_short_term_variability **46%**



Dropped the most correlated feature due to collinearity that can lead to redundant information and model overfitting.

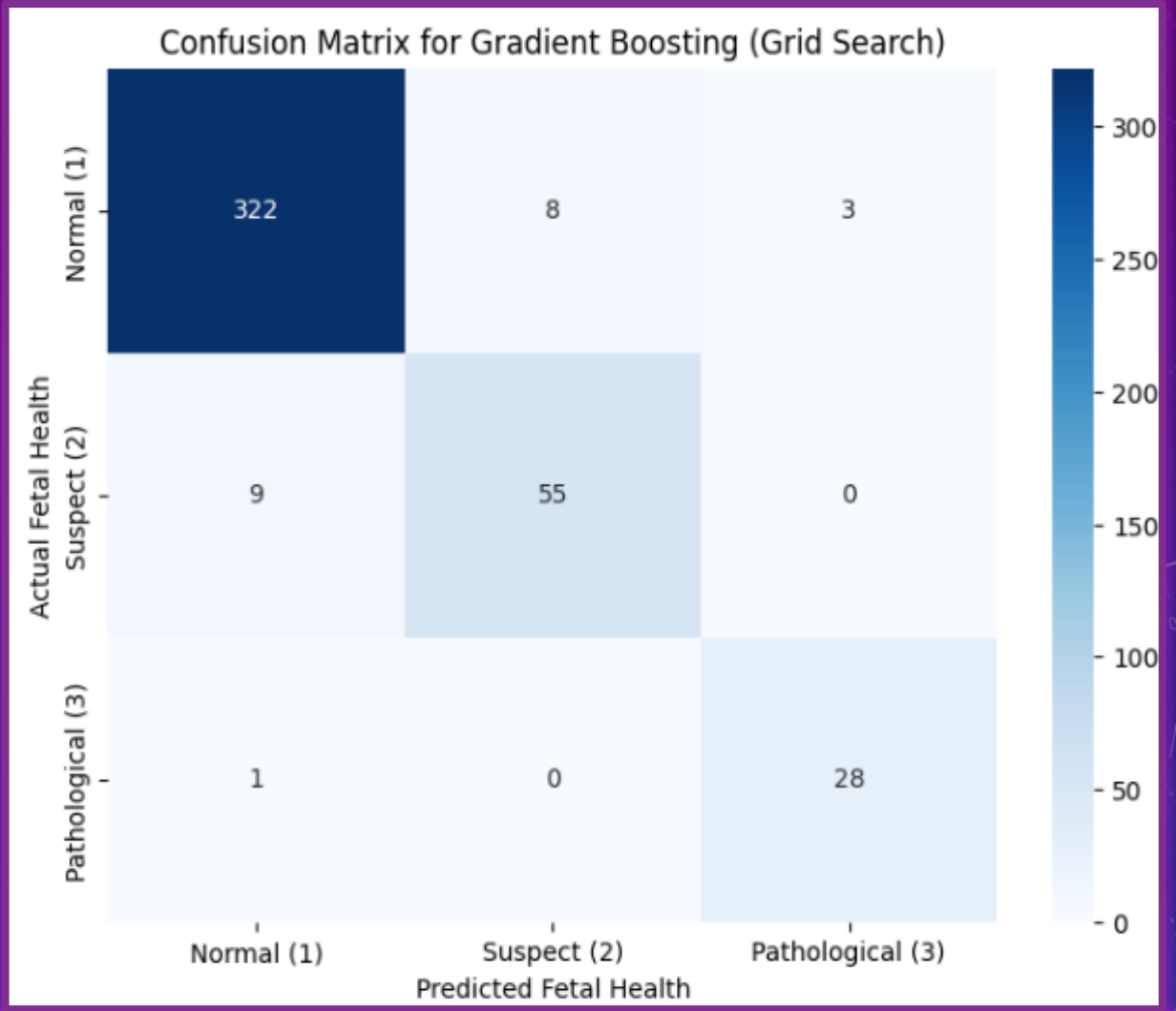
MODEL TRAINING & RESULTS

GridSearchCV tests multiple hyperparameter combinations to find the best-performing model settings, improving accuracy.

K-Folds Cross-Validation splits the data into k parts, training on k – 1 and validating on the remaining one in rotation, to estimate model performance reliably.

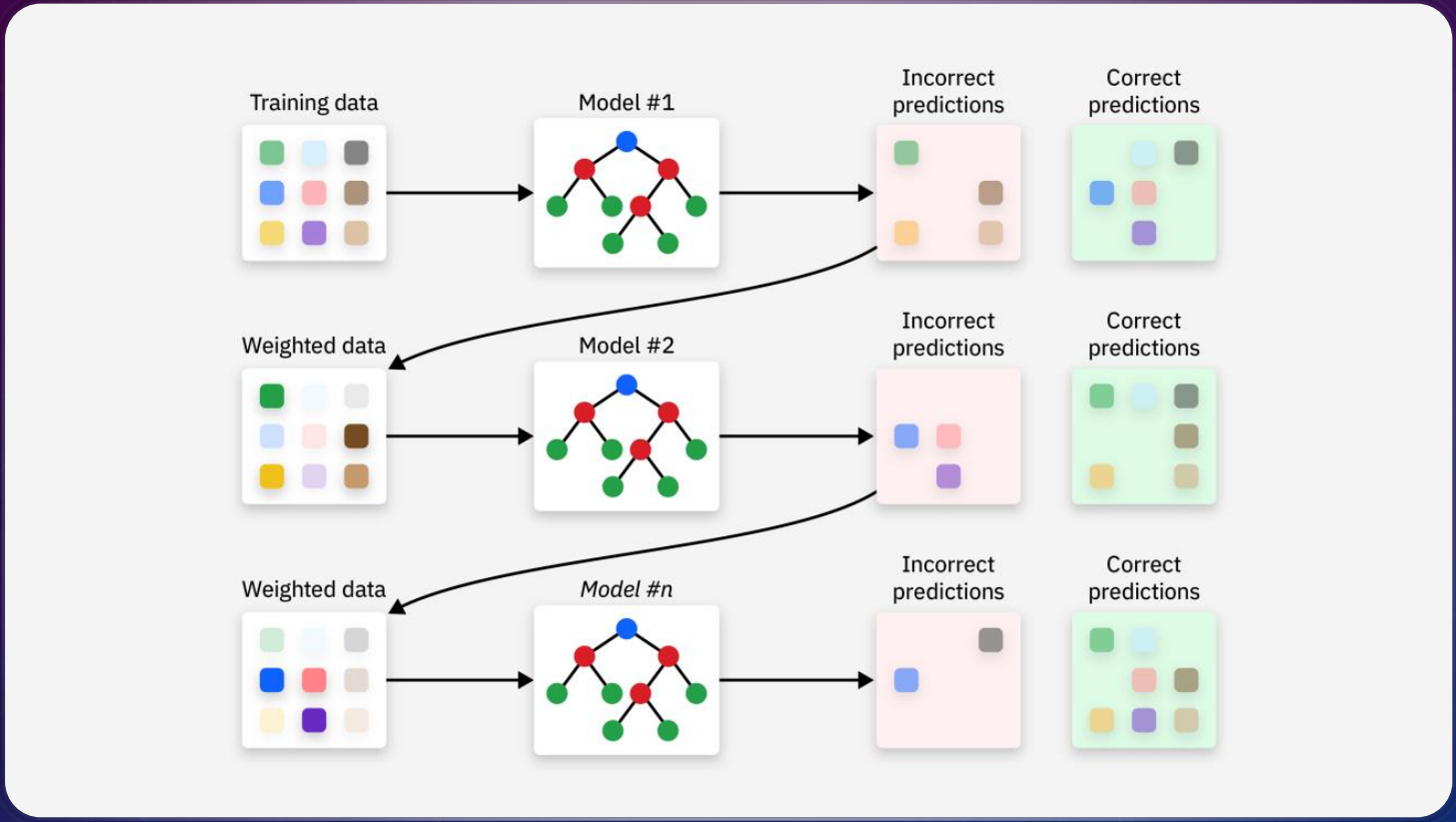
Model score for K-Fold CV = 90%

No statistically significant difference between models (all t-test p-values > 0.05).



	Accuracy	Precision (weighted)	Recall (weighted)	F1-score (weighted)
KNN (k=4)	0.899061	0.899935	0.899061	0.898674
Decision Tree	0.922535	0.924926	0.922535	0.923522
Random Forest	0.946009	0.944710	0.946009	0.944523
Gradient Boosting	0.948357	0.947965	0.948357	0.948015
Gradient Boosting (Grid Search)	0.950704	0.950790	0.950704	0.950667

GRADIENT BOOSTING IN DETAIL

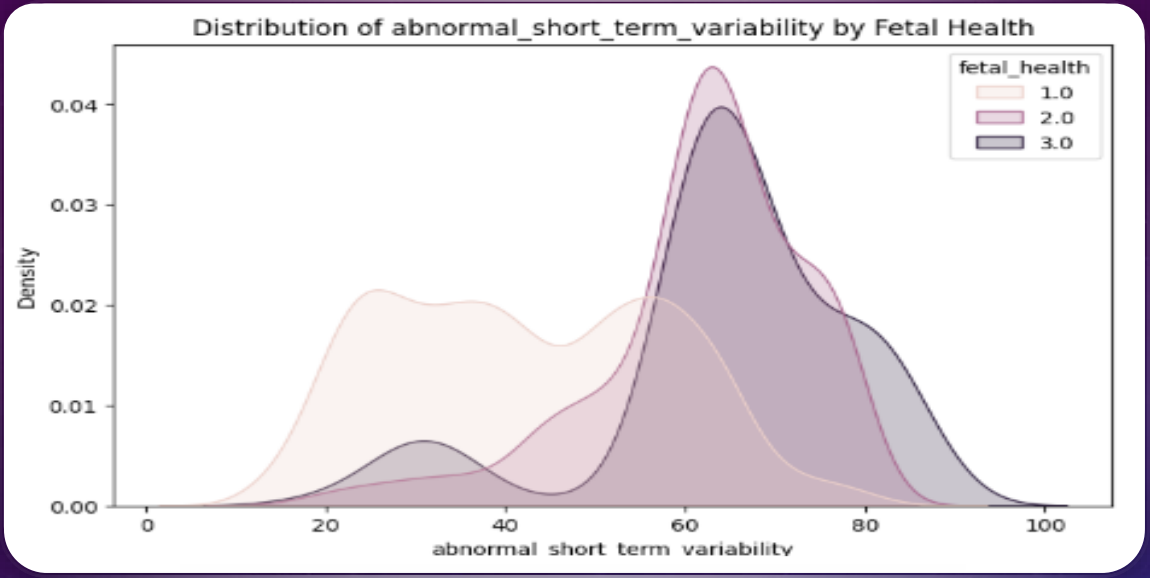


Combines many weak learners (shallow decision trees)

Gradient Boosting is an ensemble method that builds models one after another, with each new model learning from the mistakes of the previous ones to improve overall accuracy.

An Ensemble Method is a ML technique that combines multiple models to improve overall prediction accuracy by using their collective strengths and reducing individual errors.

Research Question Part #2



Gradient Boosting Classification Report (Variability Features Only):

	precision	recall	f1-score	support
1.0	0.94	0.97	0.95	333
2.0	0.87	0.75	0.81	64
3.0	0.92	0.79	0.85	29
accuracy			0.93	426
macro avg	0.91	0.84	0.87	426
weighted avg	0.93	0.93	0.93	426

Feature Importances (Variability Features Only):

abnormal_short_term_variability	0.394155
percentage_of_time_with_abnormal_long_term_variability	0.238546
mean_value_of_short_term_variability	0.188468
mean value of long term variability	0.178830

Only used columns that measured Heart Rate variability

Key Insight: HRV as a Marker of ANS Maturation

- Abnormal Short-Term Variability was the most important feature for predicting fetal health.
- Supports the hypothesis that HRV reflects autonomic nervous system development.
- HRV is controlled by the ANS, which regulates heart rate accelerations and decelerations.
- Abnormal HRV patterns indicate dysfunction or immaturity in ANS regulation.
- Thus, HRV features serve as non-invasive indicators of fetal neurological development

PART #2 CONNECTION TO SCIENTIFIC RESEARCH

Study Reference: Schneider et al. (2009) Physiological Measurement

Investigated how a baby's nervous system develops before birth by analyzing HRV

- **Before 32 weeks of pregnancy:** The parasympathetic system starts to become more active. **After 32 weeks:** The sympathetic system becomes more noticeable.

- **Non-accelerating HR:** shows that the parasympathetic system is starting to develop.
- **Accelerating HR:** shows the sympathetic system is becoming more active

Conclusion: The period before 32 weeks gestation is critical for parasympathetic development.

- HRV parameters provide non-invasive biomarkers of ANS maturation.
- Supports the use of HRV analysis in assessing fetal neurological development.

CONCLUSIONS

- ***(Answer to Part #1)*** We can accurately predict fetal health using CTG features in the given dataset, showing that CTG measurements contain reliable predictive information about fetal health

- **(Answer to Part #2)** HRV serves as a marker of ANS maturation because the ANS directly controls HRV patterns. Our ML model outcomes suggest that abnormal short-term variability may be a useful indicator of fetal ANS dysfunction.

- **Future Directions:** Validate findings on larger datasets, implement the model in clinical monitoring systems, and explore additional HRV parameters for enhanced prediction accuracy

- **Impact:** This work enables earlier detection of fetal neurological issues, supports development of automated prenatal screening tools, and advances personalized care by bridging machine learning with clinical understanding of fetal development

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Reference Paper: Schneider et al. (2009). *Fetal heart rate variability reveals differential dynamics in ANS development.* Physiological Measurement. PubMed