Can machine learning models help us read CTG results?

CAPSTONE THREE PRESENTATION

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Did you know...

Every day, 800
women die from
preventable causes
related to
pregnancy and
childbirth around
the world.

In 2021, 5 million deaths in children under the age of 5 was recorded, mostly from preventable and treatable causes.

United Nations World Health Goals

Reduce maternal mortality Reduce newborn and under-5 mortality

Cardiotocography

Simple

Cost Effective Improve Pregnancy Outcomes

Human Error in CTG Interpretation

Some measurements can't be quickly calculated

- High variability between interpretations
- Inaccurate interpretations lead to negative outcomes.

Negative Outcomes

- C-sections
- Metabolic acidosis (build up of acid in the body)
- Preterm delivery
- Low birthweight
- poor health metrics
- ► A stay in NICU

Computerized Analysis of CTG Results

Can a computer accurately classify fetal health?

Studies show promising results!

2016 Study by Ignatov

- Goal: Evaluate the effectiveness of Computerized CTG compared to conventional CTG
- ► Sample size: N =720 women in active labor
- 360 women randomly assigned to computerized CTG, remaining giving conventional CTG
- All newborns from both groups were assessed in the same way.

Their Conclusion:

The incidence of adverse perinatal outcomes was lower among women who were monitored by computerized CTG.

This ML project

Can we build a machine learning model to predict fetal health?

Our Data:

- ▶ 2126 examples of CTG readings.
- ► Features extracted from CTG outputs and labeled by professionals
- ▶ 3 Categories:
 - 1. Normal
 - 2. Suspect
 - 3. Pathological

Our Goals:

Greater than 80% accuracy

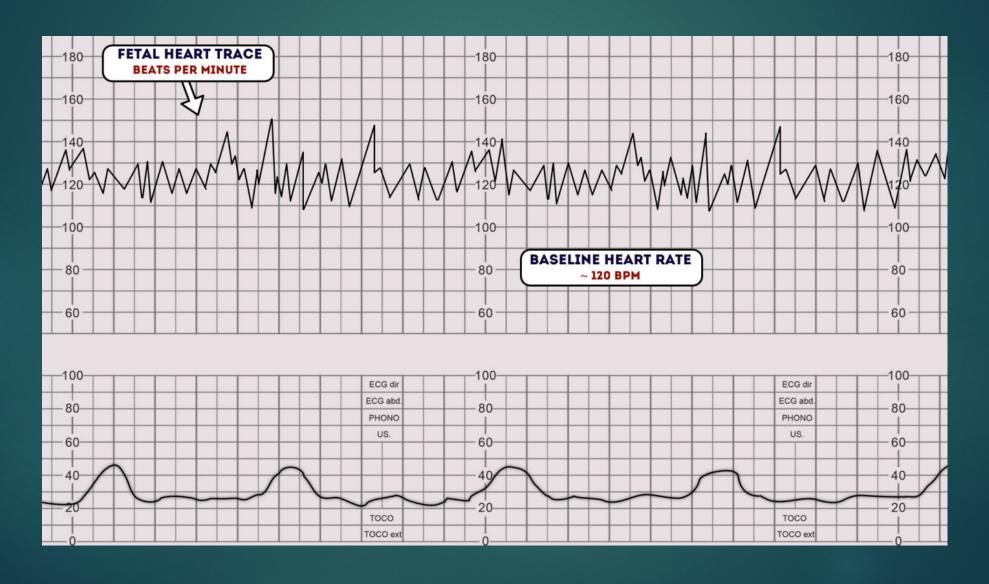
Class 3 recall greater than 90%

Identify most important features

Our Features

- ▶ We have 22 columns
- ▶ First 11 columns are measurements taken from the CTG
- ▶ They deal with:
 - 1. Baseline Rate
 - 2. Variability
 - 3. Accelerations
 - 4. Decelerations

An example of a CTG printout



Baseline Rate

The average fetal heart rate

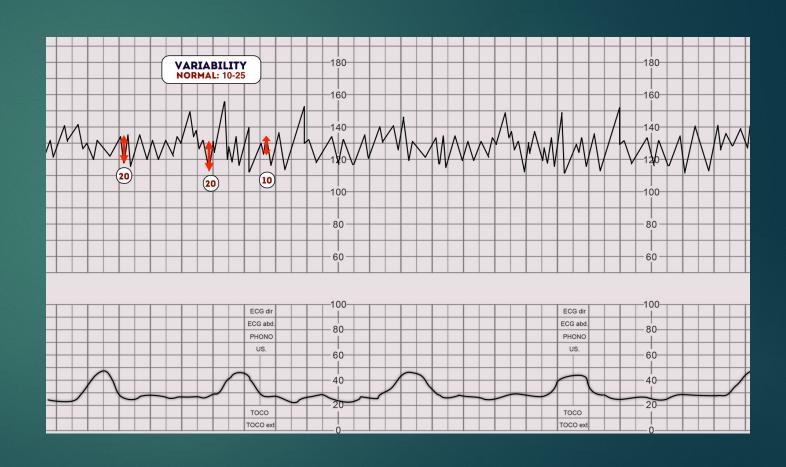
Normal is between 110 to 160 bmp

Variability

Fluctuations in fetal heart rate

Normal is between 5 to 25 bmp

Measured from peak to following trough

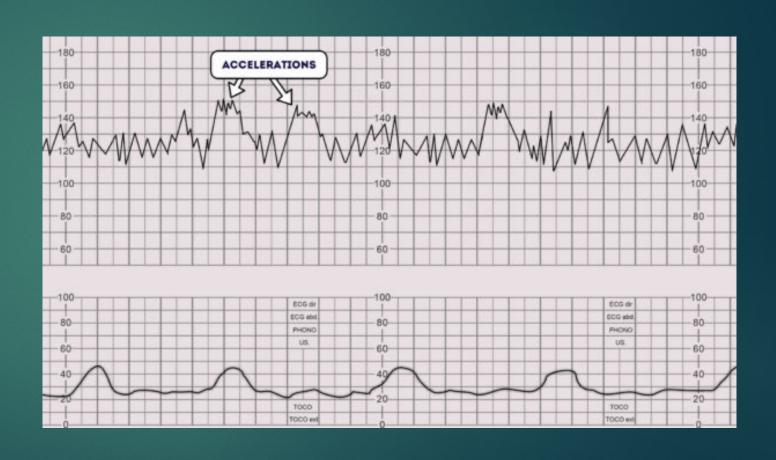


Accelerations

Short increases of FHR above the baseline

Above the baseline by >15 bpm for > 15 sec

A sign of good fetal health



Decelerations

Short decreases of FHR below the baseline

Below the baseline by 15 bpm for 15 sec

Many different types



Histogram of Fetal Heart Rate

- Remaining feature columns extracted from a histogram of FHR
- Measures things like:
 - ▶ Mean
 - Median
 - Mode
 - Number of Peaks
 - Width
 - Max
 - Min

Positive Correlation with Target

- Prolonged deceleration
- Abnormal short-term variability
- Percentage of time with abnormal long-term variability

Negative Correlation with Target

- Accelerations
- Uterine Contractions
- Mean Value of Long-Term Variability

Early Models

Dummy Model for Baseline

- Just predict the most common class every time
- ► Accuracy 78%
- ► Class 3 Recall of 0

First Real Models

▶ Decision Trees

Accuracy: 89%

Class 3 Recall: 77%

Random Forest

Accuracy: 93%

Class 3 Recall: 89%

Best Multiclass Model Gradient Boosting Classifier

Accuracy of 94% Over 20% increase!

Class 3 Recall of 93%

Assessing the Models

- ▶ All the models struggled the most with class 2 Suspect.
- Would combining Classes 2 and 3 into one category help?
- ▶ New Categories:
 - ▶ 1 = Suspect or Pathological
 - ▶ 0 = Normal

Best Binary Model

Gradient Boosting Classifier

Accuracy of 96%

Class 1 Recall of 88%

Which is Better?

- Suspect Category is hard to pick out
- Worse Case: Pathological Classified as Normal
- ▶ The Multiclass Model misclassified:
 - 2 Pathological as Normal
 - ▶ 19 Suspect as Normal
- ▶ The Binary Model misclassified 14 cases as Normal
 - ▶ We don't know if these were Suspect or Pathological

Keep all three classes

Machine Learning has a lot to offer in this area!

- With the aid of computerized CTG we could:
 - Reduce the under-5 mortality rate
 - Improve decision making strategies for labor and delivery
 - 3. Enable medical professionals to help more patients
 - 4. Decrease the occurrences of negative outcomes for newborns

Conclusion

Questions and Continued Study

- Since we are questioning the labels, it would be interesting to do a clustering analysis of the data without the labels.
- It would be very helpful to have the negative outcomes for each birth.
- How many of the Suspect cases were healthy babies?
- ► How many of these healthy class 2's did the model misclassify? In the end, who was right?
- We need more studies with larger sample sizes.