



Optimizing recovery conditions in female soccer athletes using machine learning

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Background

- VWSO Tracking
- Lack of predictive models
- Lack of external factors
- Importance of proper recovery



Related Work

- Fitness trackers and improved data collection
- Advanced biometrics
- Heart rate variability
- HRV studies
- Nonlinear models



Methods: Data

- Data from VWSO (Firstbeat trackers) & NOAA
 - 98 variables
 - 77 not relevant to this research





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- Relevant Variables
 - HRV attributes
 - Intensity
 - Miscellaneous

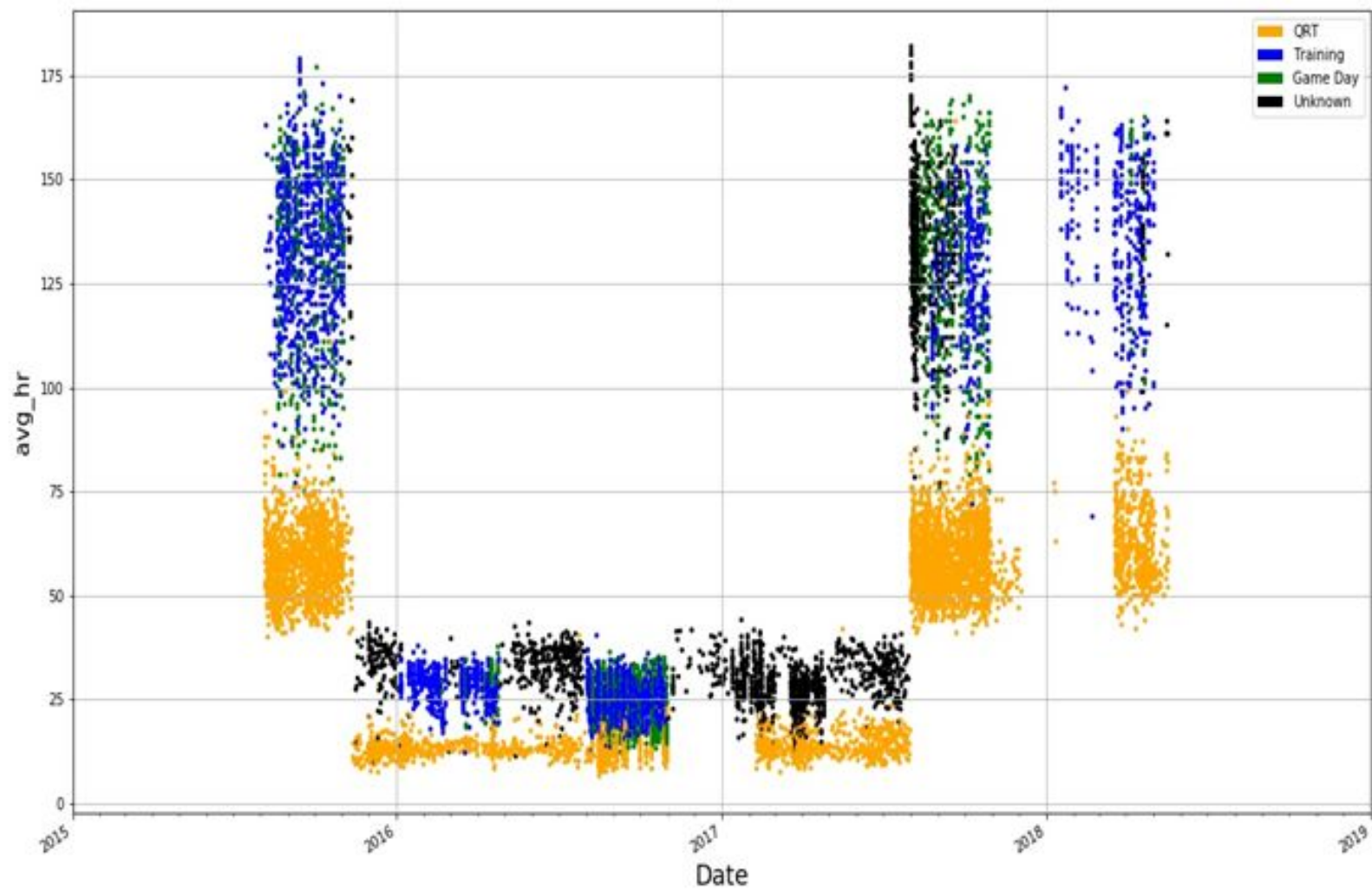




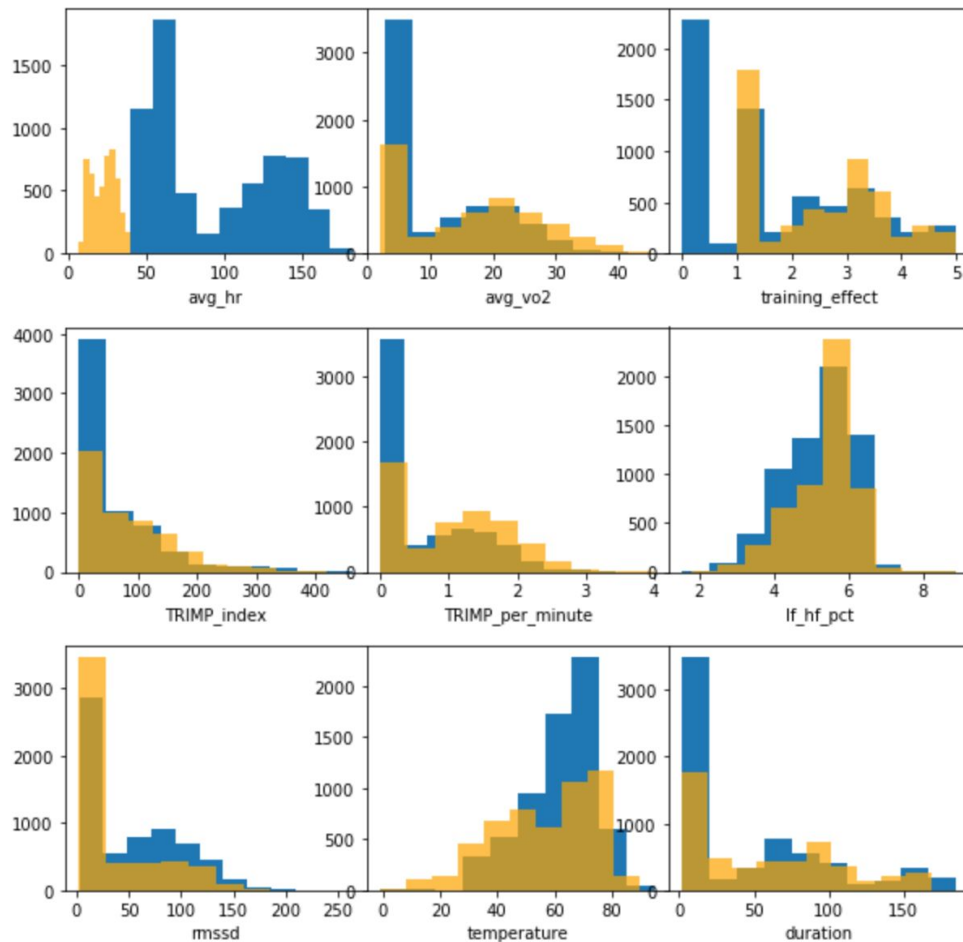
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Distributions of Attributes





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Methods: Models

- Team Model and Individual Models
- Linear Mixed Effects Model
- Random Forest Regression



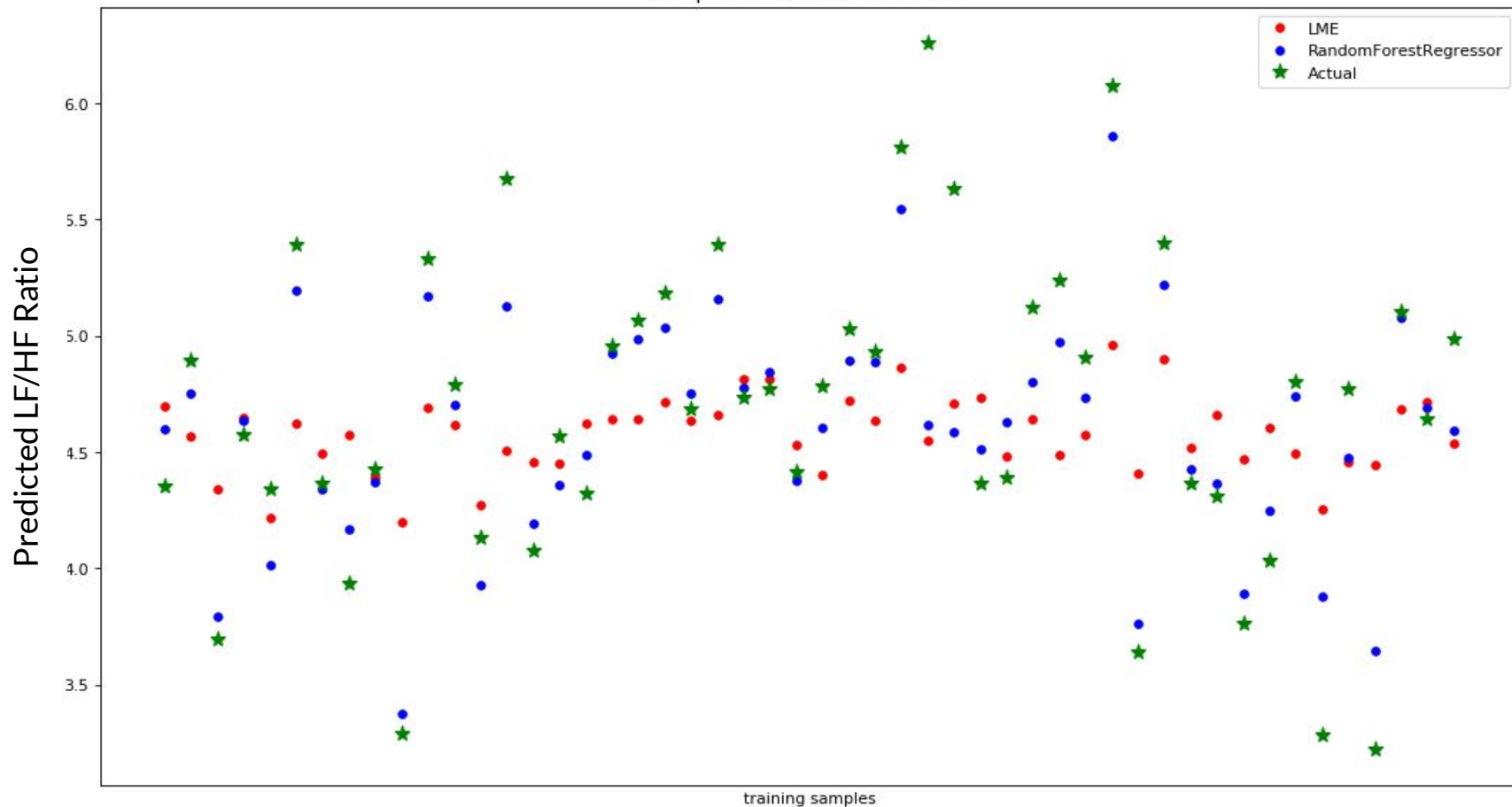
Results



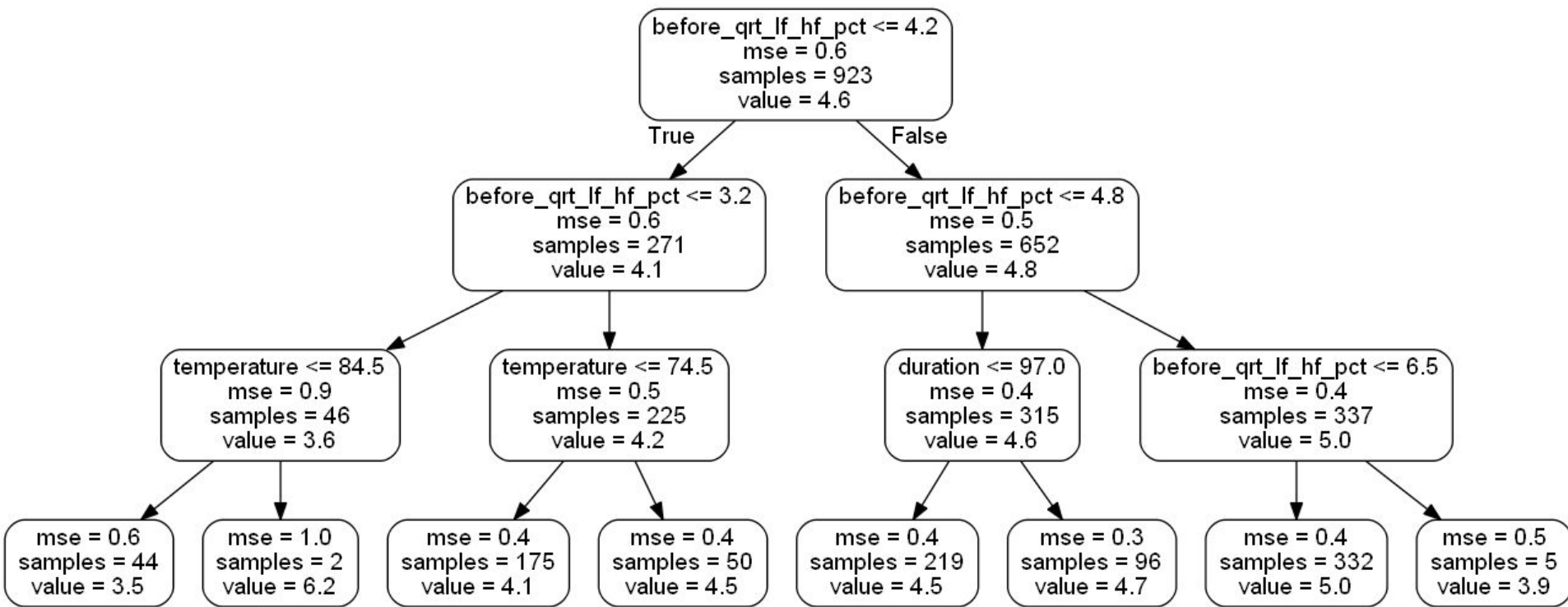
| | LME | Before/After LME | Individual RF | Team RF | Before/After RF |
|-----------|---------|------------------|---------------|----------|-----------------|
| QRT Index | 164.14% | 49.52% | 529.88% | 197.63% | 34.97% |
| QRT % | 245.59% | 458.00% | 11690.59% | 238.23% | 260.40% |
| LF | 838.31% | 560.00% | 150.96% | 4766.27% | 29.21% |
| HF | 139.66% | 120.00% | 750.30% | 218.05% | 78.37% |
| Lf/Hf % | 115.21% | 78.50% | 258.06% | 196.44% | 11.55% |
| VLF | 126.00% | 99.00% | 560.04% | 888.17% | 206.93 |
| RMSSD % | 493.54% | 268.00% | 35372.41% | 3264.01% | 10.59 |
| RMSSD | 120.77% | 99.00% | 122.32% | 506.32% | 30.01% |

Mean Absolute Percentage Error for all Models and HRV attributes

Comparison of Model Predictions



The Random Forest predicts much closer to the actual values compared to the Mixed Effects Model

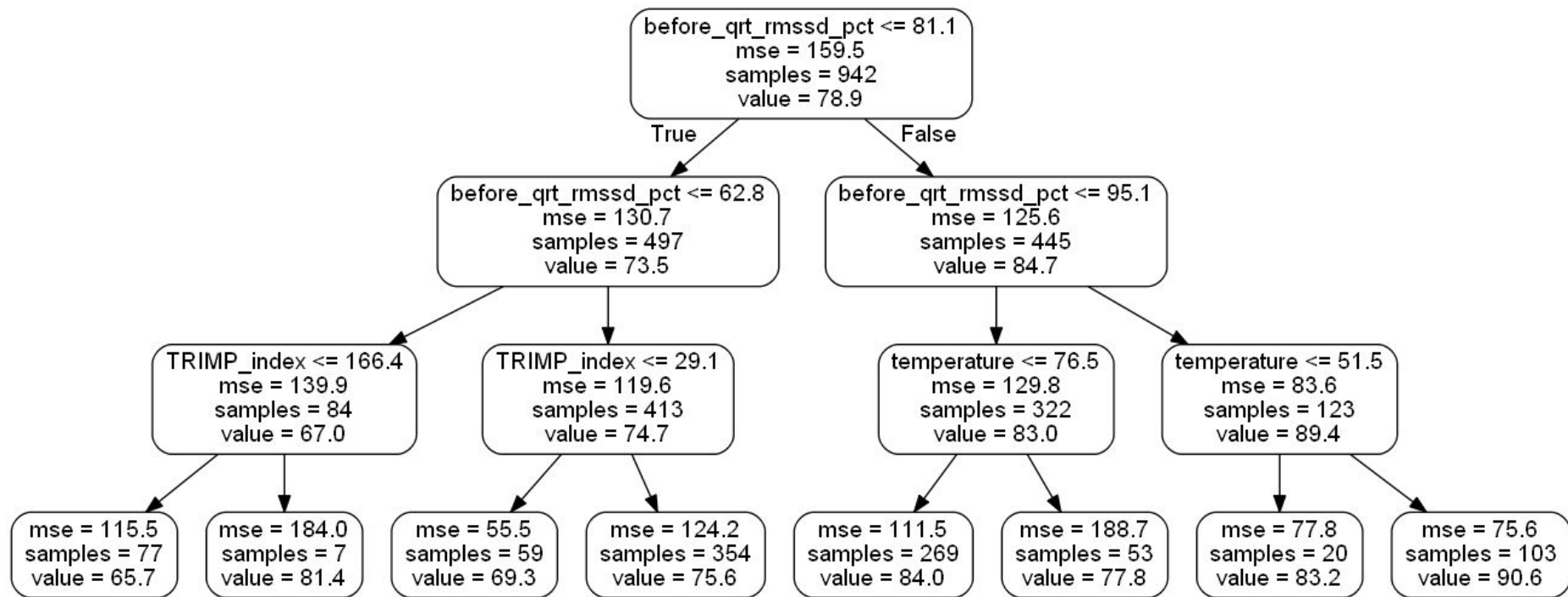


A single tree from the random forest model predicting LF/HF Ratio after an Activity



| | LME | Before/After LME | Individual RF | Team RF | Before/After RF |
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Mean Absolute Percentage Error for all Models and HRV attributes



A single tree from the random forest model predicting RMSSD % after an Activity



Results

- Linear Mixed Effects Model does not perform well
- Random Forest
 - does not perform well on individual player data or team data
 - performs better on data using former HRV measurements as a predictor



Conclusions

- Biometrics often do not perform well under linear analysis
- More individual player data is needed to perform predictions on individual data
 - If every player increased their usage of the Firstbeat tracker, this should be revisited
- It makes sense to use the pre-activity HRV measurements as this is a baseline from which the player is potentially moving from.
- This data shows that when in certain ranges for each HRV, temperature has a varying effect on post-activity HRV.
- A player in a well recovered state, subjected to high temperatures will be in a less recovered state post-activity relative if they had performed that activity at moderate temperatures
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Continued Work

- Neural Network
- Improve data quantity and quality