



Winning Edge: Texas Hold'em Data Analysis

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Background

WinningEdge™ is a strategy optimization tool for online, low-stakes Texas hold'em players. The application will provide real time probabilities and suggest action based on an ML algorithm trained using historical data. In using WinningEdge™, players will be able to make better informed decisions with the goal of winning and being profitable.



Data

- >500,000 low-stakes, heads up texas hold'em poker games purchased from 888poker.com
 - 888poker.com is one of the three largest online poker sites.
 - Data stored in >30,000 .txt files of various lengths.
- Data included in each game:
 - # of players, usernames, and who is small and big blind.
 - Game ID#
 - Time and Date
 - Chip stack sizes
 - All betting actions at each phase of game (preflop, flop, turn, river).
 - Bet amounts
 - Board cards at each phase of game. (if applicable)
 - Player cards. (if nobody folds)



Data Example:

This is an example of a single game played

- Here you can see the structure of the data given directly from 888poker.com

```
#Game No : 502745408
***** 888poker Hand History for Game 502745408 *****
$0.01/$0.02 Blinds No Limit Holdem - *** 06 06 2018 04:49:57
Table Bedford 6 Max (Real Money)
Seat 1 is the button
Total number of players : 2
Seat 1: ponte1001 ( $0.80 )
Seat 4: Bolorig888 ( $1.01 )
ponte1001 posts small blind [$0.01]
Bolorig888 posts big blind [$0.02]
** Dealing down cards **
ponte1001 calls [$0.01]
Bolorig888 checks
** Dealing flop ** [ 2c, Qh, Jd ]
Bolorig888 checks
ponte1001 checks
** Dealing turn ** [ 9h ]
Bolorig888 bets [$0.04]
ponte1001 calls [$0.04]
** Dealing river ** [ Js ]
Bolorig888 bets [$0.06]
ponte1001 calls [$0.06]
** Summary **
Bolorig888 shows [ 9c, Qs ]
ponte1001 mucks [ 9s, 6h ]
Bolorig888 collected [ $0.23 ]
```

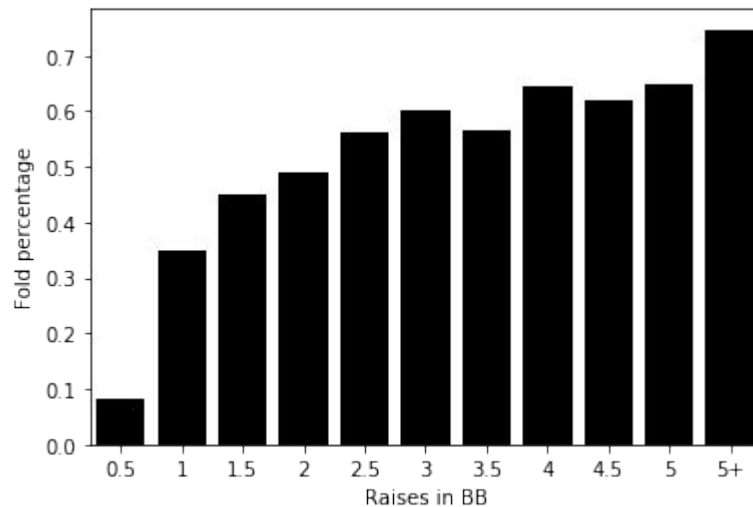
Use Case 1: Showdown Analysis

Which hands are more likely to make it to the end?



Use Case 2: Bet comparison

What size bet can make the opponent fold for the cheapest price?





Use Case 3: Catching Bluffs

You made it to the river. Opponent bets. Given the board and how the hand was played, how do you know if its a bluff?

Our first model considers only the board texture and aggression level of opponent.

57% accuracy is low.

To improve, we need better features.

```
In [3]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=40)

model = LogisticRegression()
model.fit(X_train, y_train)

y_pred = model.predict(X_test)

accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy * 100:.2f}%")

precision = precision_score(y_test, y_pred)
print(f"Precision: {precision * 100:.2f}%")

recall = recall_score(y_test, y_pred)
print(f"Recall: {recall * 100:.2f}%")

Accuracy: 57.31%
Precision: 52.57%
Recall: 31.56%
```

```
In [4]: coefficients = model.coef_
intercept = model.intercept_

print("Coefficients:", coefficients)
print("Intercept:", intercept)

Coefficients: [[ 0.2130902  0.21575425  0.00345047 -0.33398528 -0.1852743 -0.56403836
 -0.23594029  0.11906768]]
Intercept: [-0.15363636]
```

Model: Under the hood

Inputs:



board



bets



winners

Metrics:

Flushyness

Straightyness

Aggression

Bluffs

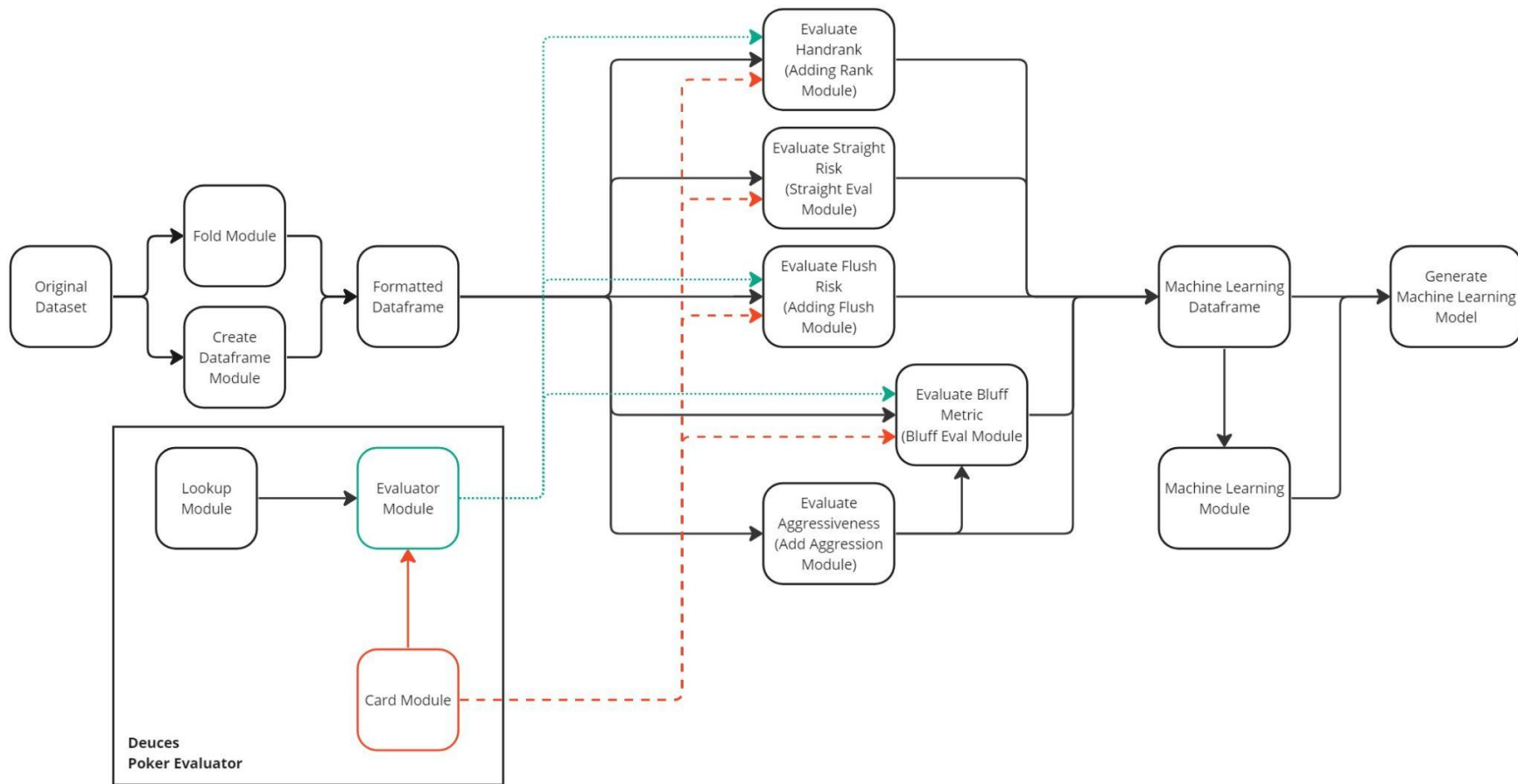
Model:

Features

Outcomes



Design:



Evaluator Demo

```
%cd ..
from winedge import Card, Evaluator

# for debugging purposes, also include ```Lookup```
```

\\ws1.localhost\Ubuntu-22.04\home\dheino13\WinningEdge

Input hands and board

Provide hands information of you and your opponent, as well as the community cards.

Name the cards in the following way:

- For ranks, use ['23456789TJQKA'] .
- For suits, use ['shdc']
- For example, ten of heart would be named ['Th'], whereas eight of spade is denoted as ['8s']

```
hand1 = [
    Card.new('Ts'),
    Card.new('Ac')
]

hand2 = [
    Card.new('2c'),
    Card.new('Js')
]

board = [
    Card.new('2h'),
    Card.new('2s'),
    Card.new('3c'),
    Card.new('Kh'),
    Card.new('Qc')
]
```

```
# Initialize evaluator. This step calls the lookup function and generate
# the LookupTable for hand ranks.
evaluator = Evaluator()
```

```
# evaluator.evaluate(hand, board) provide hand ranks
# other functions provide different ways of representing hand strengths
```

```
rankclass1 = evaluator.class_to_string(evaluator.get_rank_class(evaluator.evaluate(hand1, board)))
rankperc1 = evaluator.get_rank_percentage(evaluator.evaluate(hand1, board))
```

```
rankclass2 = evaluator.class_to_string(evaluator.get_rank_class(evaluator.evaluate(hand2, board)))
rankperc2 = evaluator.get_rank_percentage(evaluator.evaluate(hand2, board))
```

```
print("Rank class for your hand is: %s" % rankclass1)
print("The hand strength is: %f" % rankperc1)

print("Rank class for your opponent's hand is: %s" % rankclass2)
print("The hand strength is: %f" % rankperc2)

if rankperc1 > rankperc2:
    print('\nYou won!')

elif rankperc1 < rankperc2:
    print('\nYou lost!')

else:
    print("\nIt's a draw!")
```

Rank class for your hand is: Straight
The hand strength is: 0.785580
Rank class for your opponent's hand is: Full House
The hand strength is: 0.957920

You lost!



Repository Structure:

- Dataframe directory
 - Formatted dataframe
 - Create dataframe module
- ML directory
 - ML dataframes
 - ML module
- Notebooks directory
 - Module prototypes
- Winedge directory
 - Data and dataframe analysis modules

```
../WinningEdge/  
├── LICENSE  
├── README.md  
├── dataframe  
│   ├── convert_data_to_dataframe.ipynb  
│   ├── create_dataframe.py  
│   ├── poker_dataframe_bugfixed.pkl  
│   └── poker_dataframe_with_agg_columns.pkl  
├── doc  
│   ├── Component Specification.pdf  
│   ├── Functional Specifications.pdf  
│   ├── Technology Review Slides.pdf  
│   └── use-case.md  
├── environment.yml  
├── examples  
│   ├── README.md  
│   └── eval_comp_demo.ipynb  
├── ml  
│   ├── README.txt  
│   ├── df_filtered_w_flush.pkl  
│   ├── df_filtered_w_flush_strength.pkl  
│   ├── make_ml_df.ipynb  
│   ├── make_regression_model.ipynb  
│   ├── needs_labels.pkl  
│   └── ready_to_model.pkl  
├── notebooks  
│   ├── AddingFlush.ipynb  
│   ├── AddingRank.ipynb  
│   ├── WIP.ipynb  
│   ├── bluff_metric_prototype.ipynb  
│   ├── straight_risk_prototype.ipynb  
│   └── testing_create_dataframe.ipynb  
├── setup.py  
└── winedge  
    ├── Folded.ipynb  
    ├── __init__.py  
    ├── __pycache__  
    ├── add_aggression_column.py  
    ├── add_fold_column.py  
    ├── adding_flush.py  
    ├── adding_rank.py  
    ├── bluff_eval.py  
    ├── card.py  
    ├── evaluator.py  
    ├── lookup.py  
    ├── straight_risk_eval.py  
    └── tests.py
```

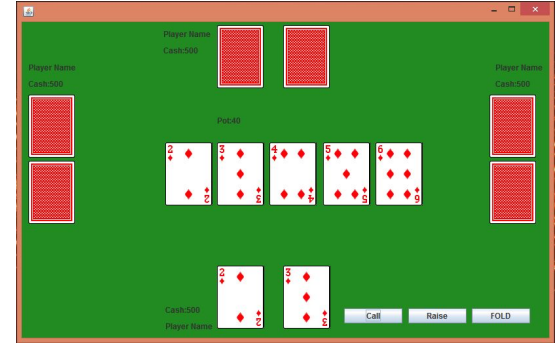


Lessons Learned:

- Understanding limitations and possible biases of data prior to modelling
 - Player cards only visible for completed games introduces heavy bias to ML model
- Importance of a comprehensive dataset
- Benefits of Modular Design
 - Breaking problem into smaller parts (Dataframe, Evaluator, etc.)
- Importance of complete data
 - Handling missing / invalid data

Future Work:

- Refine ML algorithm to increase accuracy
- Purchase more comprehensive dataset to train ML and eliminate biases
- Implement **GUI** to utilize software in real time while gaming
- Implement continuous training so training dataset can be expanded in real time while in use
- Expand scope of software to higher stakes game and >2 players





Thanks!