UFC Fight Data Analysis: Storyboard

1. Introduction to the Data and Case

In this analysis, we delve into UFC (Ultimate Fighting Championship) fight data to uncover patterns and insights that can inform fighter strategies and enhance our understanding of the sport. The dataset, titled "UFC Final Data," includes a range of metrics related to fight outcomes, fighter characteristics, and fight events, such as significant strikes, takedowns, submission attempts, and more.

Objective: Our goal is to explore the dataset, validate specific hypotheses related to fight outcomes, and provide actionable insights based on the data analysis. We will focus on analyzing fighter stances, striking metrics, and submission effectiveness to determine their impact on fight success.

2. Exploratory Data Analysis (EDA)

We began by conducting an Exploratory Data Analysis (EDA) to understand the overall structure and distribution of the data. Key steps included:

- Data Cleaning and Preparation: We removed any irrelevant historical events, ensured no forecasted data was included, and addressed any inconsistencies in the dataset.
- **Visualization of Fighter Stances:** We visualized the distribution of different fighter stances (Orthodox, Southpaw, Switch, etc.) to understand their prevalence and how they might relate to fight outcomes.
- Initial Research Questions: Our exploration was guided by questions like:
 - Which fighter stance is most common among top fighters?
 - Is there a correlation between specific fighting techniques (e.g., significant strikes, takedowns) and fight victories?
 - o How does a fighter's geographical location affect their success in the UFC?

3. Hypothesis and Key Question

Hypothesis 1: Winning fighters have a significantly higher average number of significant strikes landed compared to losing fighters.

Research Question: "Does the fighter's stance (Orthodox, Southpaw, etc.) significantly impact the likelihood of winning a fight?"

This question became the pivot of our analysis, guiding the subsequent steps as we explored various aspects of the dataset.

4. Analyses and Results

- **Visualization of Top Fighters:** We identified and visualized the top 30 UFC fighters based on the maximum number of wins. This helped us isolate the most successful fighters for more detailed analysis.
- Correlation Between Wins and Strike Metrics: We plotted wins against various striking metrics (e.g., significant strikes, ground strikes) to see if higher strike counts correlate with more wins. Interestingly, while a positive correlation was observed in some cases, it wasn't universally strong across all strike types.
- **Submission Success Analysis:** We focused on the sum of wins by submission and compared it with average takedowns landed and attempted. This analysis revealed that fighters with higher takedown success rates often had more submission wins, particularly for Southpaw fighters.

Interpretation: These analyses showed that while striking is important, submission effectiveness (often linked with successful takedowns) is a critical factor in a fighter's success. The stance also plays a role, with Southpaw fighters showing distinct patterns in submission success.

5. Summary and Insights

The analyses confirm that a fighter's stance, striking accuracy, and submission effectiveness are all significant factors in determining fight outcomes. Specifically:

- **Fighter Stance:** Orthodox fighters are more common, but Southpaw fighters have unique advantages in submission scenarios.
- **Striking vs. Submissions:** While striking is a crucial aspect of the fight, submissions often turn the tide, especially for fighters who excel in takedowns.

These insights are valuable for coaches, fighters, and analysts looking to optimize fight strategies and predict outcomes.

6. Next Steps for Further Analysis

- **Predictive Modeling:** Build predictive models to forecast fight outcomes based on the identified key metrics (strikes, submissions, takedowns).
- **Temporal Analysis:** Analyze how the importance of different metrics (e.g., submissions, significant strikes) has evolved over time.
- **Deeper Stance Analysis:** Conduct a more granular analysis of stance effectiveness across different weight classes and fight locations.

7. Limitations

- **Data Completeness:** Some historical data might be incomplete, potentially affecting the accuracy of our analyses.
- **Stance Categories:** The dataset's classification of stances might oversimplify fighters' complex and dynamic techniques during a fight.
- External Factors: Other factors not included in the dataset, such as fighter training, coaching, and psychological readiness, could also significantly influence fight outcomes.

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