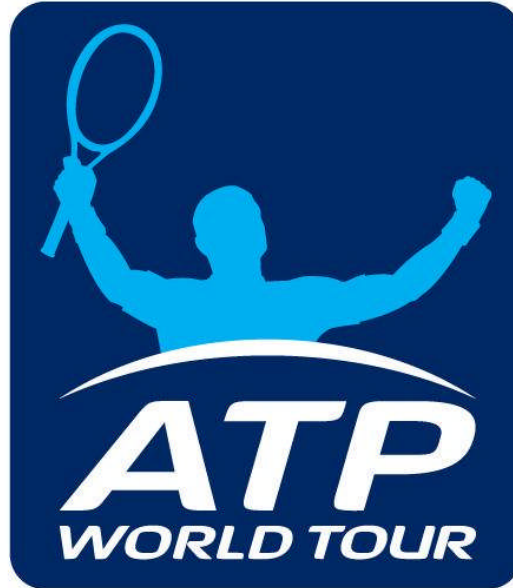

ATP 2012 – 2014 Upset Visualization

July 17, 2016



Agenda

- Context and Overview
- Odds Data Visualization
- Charted Match Statistics Visualization
- Next Steps

Agenda

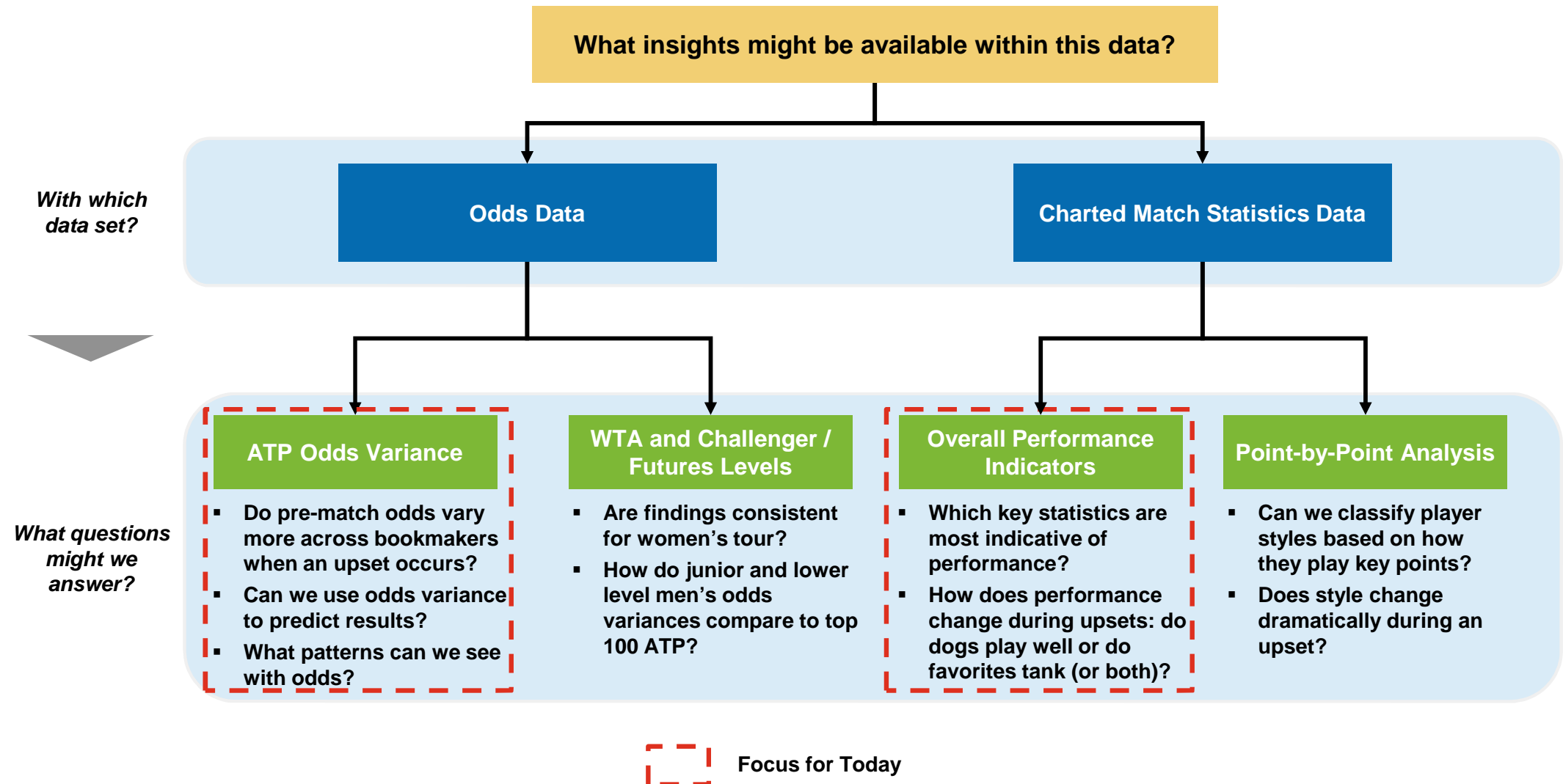
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Professional tennis is unique in that even despite the dominance of a few select competitors, major upsets do occur regularly

Analysis Context

- Relevant data sets are made available that represent two major categories of tennis match analysis:
 - **Odds**: published at www.tennis-data.co.uk, displays detailed match results and pre-match odds given by five major bookmakers on the day of each match
 - **Charted Statistics**: published by Jeff Sackmann at github.com/JeffSackmann/tennis_atp, a number of csv files are hosted with extremely detailed, point-by-point data for select matches
- I merged these disparate data sets for the highest level of professional men's tennis only (i.e. ATP not including challenger or future's matches) over the 2012, 2013, and 2014 seasons
- An upset is defined in this analysis as a completed match (i.e. player did not retire or walkover) in which the player with higher average odds defeated his opponent
- Visualization of this data intends to examine:
 - Whether any incongruities exist amongst odds makers that might predict when upsets are likely to occur
 - Frequency and severity of upsets
 - Performance by favorites and underdogs across key metrics during average, upset, and non-upset matches

There are a number of interesting questions that might be addressed with these data sets: today we will focus on two areas



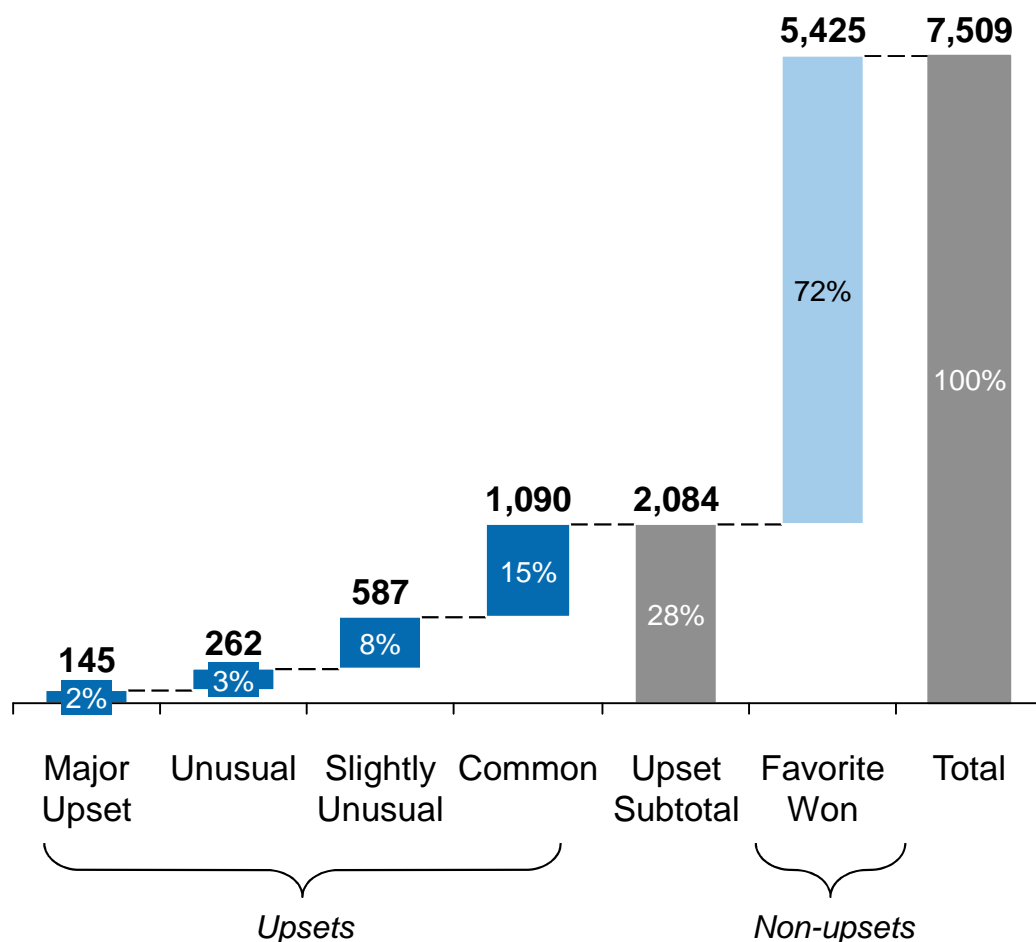
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Player odds help us understand who is truly expected to win each match; favorites tend to win approximately 72% of the time

Match Waterfall: Upset vs Non-upset Frequency

(ATP Matches from 2012-2014, n = 7,509)



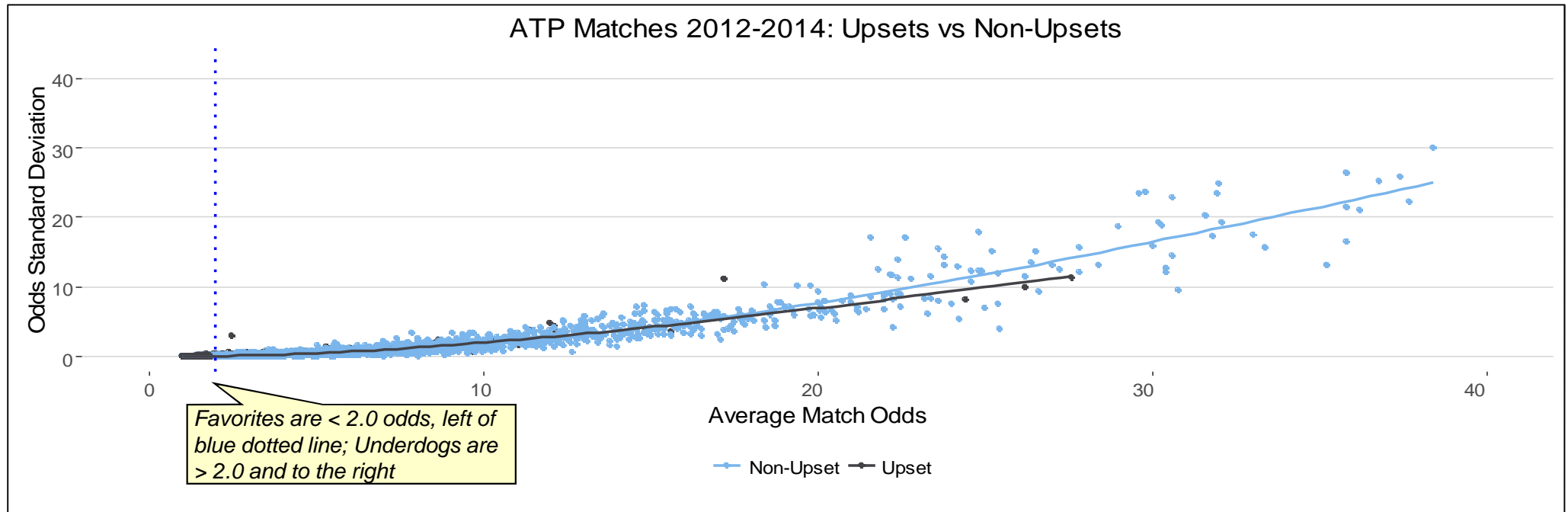
Discussion

- According to odds data, favorites win ~72% of their matches
- Remaining 28% represents underdog victories, with only 5% as either unusual or a major upset

Odds Background / Methodology

- Odds are represented in British format with favorites defined as < 2.0 and underdogs > 2.0
- This means if you were to bet \$100 on a favorite with 1.9 odds and won, you would receive \$190 from the bookmaker (original \$100 plus \$90 profit)
- Consequently, the same \$100 bet on a 3.0 underdog who won would pay \$300 (including \$200 profit)
- Assignment of players into groups is based on their average odds:
 - **Favorites:** “Slight Favorite” < 2.0, “Clear Favorite” < 1.85, “Heavy Favorite” < 1.4, and “Sure Thing” < 1.1
 - **Underdogs:** “Slight Dog” > 2.0, “Clear Dog” > 2.6, “Heavy Dog” > 3.5, “No Hope” > 5.0

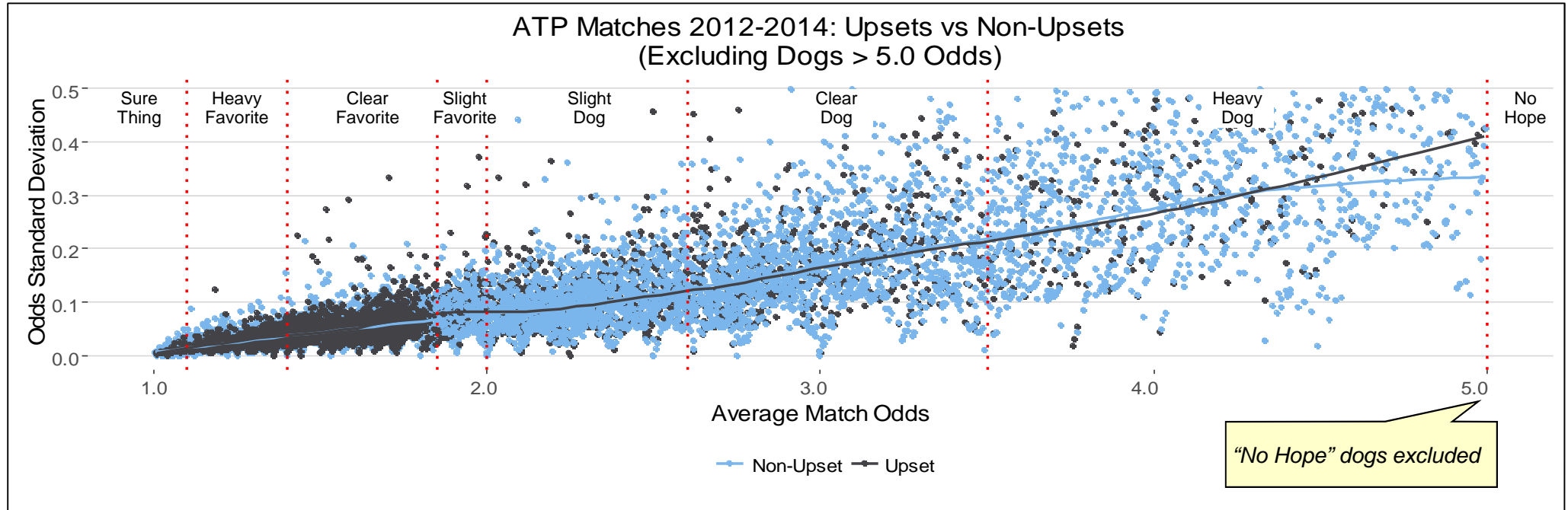
There is a clear relationship between average odds and standard deviation of odds: the bigger the underdog, the more odds vary



Discussion

- As expected, the majority of upsets (black dots) occur near the 2.0 mark, meaning the quality of players is relatively close; though there are some cases where underdogs in the 20:1 range defeated their opponents!
- In general, it seems that the amount of variance among bookmakers is much greater for major underdogs that have relatively little hope of beating the much higher ranked player
- At this level it is difficult to ascertain whether there is a meaningful difference in typical variance in upsets vs non-upsets. At first glance, it appears that the amount of variance is in line for both cases as both trend lines seem to overlap

Excluding dogs with odds > 5.0 , we can drill down and confirm that odds variance is similar for both upsets and non-upsets



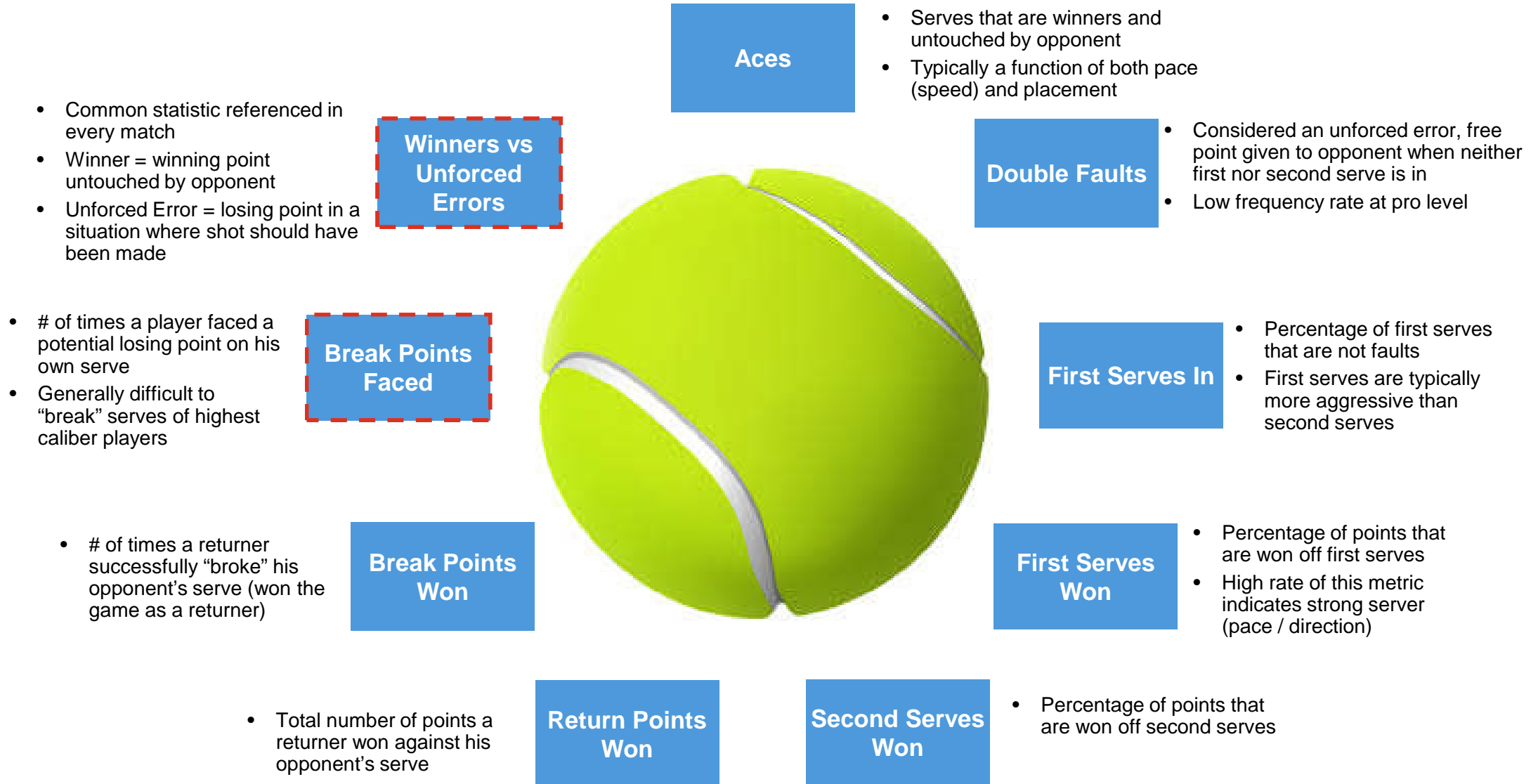
Discussion

- Within the 1.0 to 5.0 odds range, we again see much “tighter” odds for favorites across bookmakers, especially < 1.5 odds
- This could suggest that casinos put heavier focus on correctly handicapping higher probability events; that is, given limited time and resources, they may assume that odds accuracy for underdogs is less important since they are less likely to win
- While this data indicates we cannot predict upsets based on odds variance alone, it is interesting to note just how much variance exists for underdogs; bettors may use this information to carefully select bookmakers and maximize value when betting on underdogs

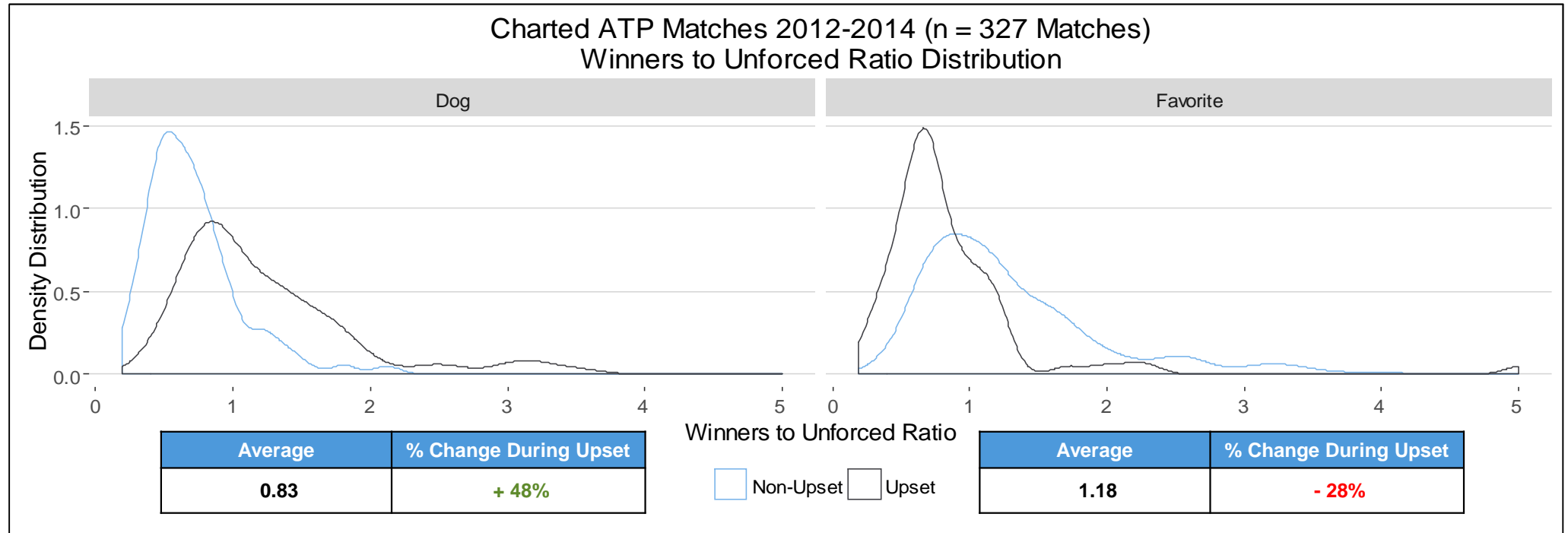
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Using the match charted data available we can track player performance against 9 key metrics



Winners to Unforced Errors ratio is the most cited statistic in tennis for good reason: it is highly correlated to match results

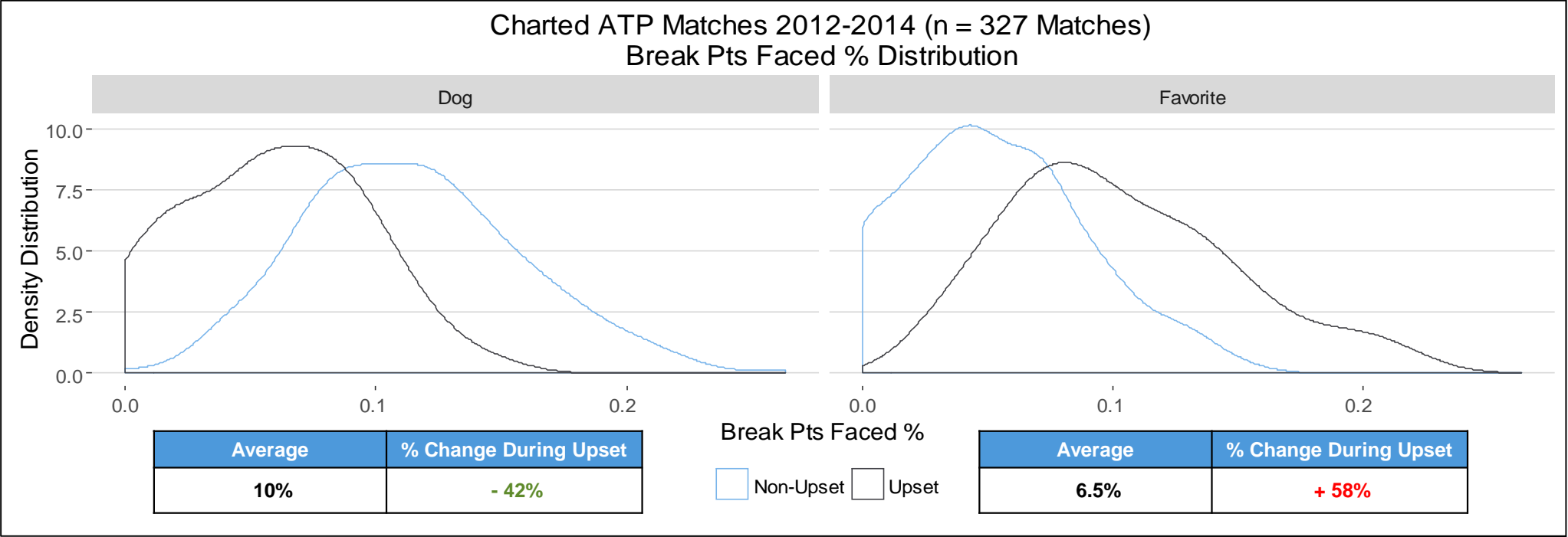


Discussion

- Winners and unforced errors constitute roughly 70% of total points played¹ and thus logically have bearing on match outcomes
- Underdogs average an unfavorable 0.83 ratio typically, though tend to improve by 48% during upsets
- This statistic would support a hypothesis that underdog performance increases more than favorite performance decreases during upsets; i.e., that the matches weren't just "bad days" for the higher ranked players

1) "A closer look at the winner-unforced error ratio" tennisabstract.com

Interestingly, break points faced is a more telling upset statistic than break points saved



Discussion

- In this case, both favorites and underdogs see dramatic performance changes during upsets, specifically to their serves
- Underdogs tend to face 42% fewer break points while favorites see 58% more
- This underscores the fact that a strong service game is critical at the pro level and can make a world of difference
- It is also generally understood that service games are more tiring for the serving player, so fewer break points faced suggests more efficient service games, which could help the underdog maintain stamina

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A fount of interesting data remains to be analyzed with time permitting

Next Steps

- Additional analyses / hypotheses that could be tested include:
 - Women's / challengers / futures tours (repeat existing visualizations)
 - Bookmaker-specific analysis (e.g., do certain books predict upsets more frequently?)
 - Point-by-Point analysis (e.g., has underdog recently changed tactics that could help us predict victory?)
 - Prior year analysis (data exists back to 1968)
 - Many more

- Questions?