

**Swinging in Sync: How Batter Mechanics Correlate with Offensive Performance from the
Individual to the Team Level**

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Introduction

Abstract

We leverage newly released Statcast metrics—specifically, bat speed and swing length—to rigorously evaluate the relationship between variability in batter mechanics and offensive productivity. Focusing on both individual and game-level analyses, we examine how variability in these swing metrics correlates with advanced offensive measures such as xwOBA. Our findings indicate that lower variability in swing length is significantly associated with higher individual performance, highlighting a possible importance of consistency in this aspect of mechanics. At the team level, however, variability in bat speed appears to exert a positive influence on daily offensive outcomes, suggesting that adaptability on a game-to-game basis in this metric may be beneficial. Using linear and mixed-effects modeling, we assess the impact of mechanical variability while accounting for team-specific factors, uncovering robust correlations that offer insights into optimizing lineup construction and player development strategies. While these results are promising, further research is needed to disentangle the causal relationships and explore how factors like pitcher matchups, situational context, and biomechanical constraints shape the connection between variability and performance.

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Overview

Recent advances in Statcast technology have allowed for the direct measurement of bat speed and swing length on pitches where the batter swings. These new metrics open the door to exploring questions that were previously left to inference, scouting reports, or assumptions about mechanics. Specifically, we have become interested in the idea of swing consistency: How much do a batter's bat speed and swing length deviate from pitch to pitch or from game to game? How do differing swing profiles on a roster correlate with run generation during a game?

This paper addresses three connected but distinct angles on these new metrics:

Individual-Level Swing Diversity: Do stable or volatile swing metrics for a single batter over a season correlate with that batter's offensive success, measured through wOBA or xwOBA?

Game-Level Team Swing Diversity: Does a club whose hitters share similar or very different swing metrics on a given day see any advantage in that day's run production?

Our analysis aims not to prove a direct cause-and-effect relationship but rather to uncover whether correlations exist that might guide roster construction, daily lineup decisions, or player development. Given that these bat speed and swing length metrics are new, we expect some initial findings to be exploratory, setting the stage for deeper or more controlled studies in the future.

Fundamental Questions Of Our Study

- I. **Individual:** Is consistency associated with higher personal performance once I factor out average skill?
- II. **Game:** On a day-to-day basis, does having a consistent or varied approach correlate with daily offensive output?

Methodology

Below, we describe the major steps involved in each portion of the analysis. The logical flow starts at the individual hitter level, and moves into game-level analyses.

Individual-Level Analysis

In this analysis, we aimed to examine whether players who maintained a more consistent swing (in terms of swing length and bat speed) tended to exhibit higher offensive performance, proxied by wOBA and xwOBA. We began by loading our pitch-level Statcast dataset and relevant libraries in R. We then cleaned and prepared the data, generating a unique plate appearance identifier and assigning each batter to the appropriate team based on whether the inning was top or bottom. For each player, we computed their season-long average (mean) and variability (standard deviation) of swing length and bat speed. By removing players with insufficient data or missing standard deviations, we ensured that we only focused on individuals with a reliable sample of swings. Next, we defined the “final pitch” of each at-bat to capture the at-bat’s ultimate outcome and aggregated each player’s expected wOBA (xwOBA). We merged these performance metrics with the consistency measures (mean and standard deviation of swing mechanics) to create a comprehensive dataset for analysis.

Subsequently, we visualized the relationship between each metric of variability and xwOBA to observe whether greater variability correlated with lower performance. We then built linear models regressing xwOBA on mean and standard deviation of swing length (and bat speed) to assess whether consistency had explanatory power once skill level (mean swing metrics) was accounted for. We checked for multicollinearity using correlation matrices and variance inflation factors (VIF), discovering only moderate correlations. Given the presence of some overlap in our variability measures, we performed Principal Components Analysis (PCA) on the standardized variables and extracted a single “mechanical variability” component that captured the common elements of variability in swing length and bat speed. Finally, we incorporated this principal component into a multivariate linear regression to see if it could successfully stand in for the separate variability measures.

Game-Level Analysis

This second analysis shifted from player-level season-long metrics to game-by-game team-level measures, exploring whether day-to-day consistency or diversity in swing mechanics (aggregated across all hitters on a team) related to that team's offensive output for a particular game. We again started by loading the same Statcast pitch dataset and preparing it in R, assigning each row to a unique plate appearance ID and identifying which team was batting. We then calculated, for each batter in each game, various statistics such as average (mean) and median bat speed and swing length, as well as the total number of plate appearances for each batter. Leveraging these per-batter game summaries, and specifically focusing on median statistics per-batter in a game due to its resistance to outliers like check swings, we used weighted means and weighted standard deviations—weighted by each batter's plate appearances in that game—to compute team-level “diversity” or “variability” in swing mechanics for every single game.

With these team-level aggregates in hand, we merged them with wOBA and xwOBA to generate a master dataset that captured both offensive performance and variability in approach. We filtered out incomplete or insufficient data (e.g., where there were zero swings or no standard deviations). We visualized possible relationships between each type of swing mechanic variability (e.g., SD of median bat speed, SD of median swing length) and the team's daily xwOBA using scatter plots, color-coded by team, only showing NL East teams as to not clutter the graph. Finally, we evaluated multivariate linear regression models to account for multiple predictors simultaneously (e.g., mean and variability of bat speed and swing length) and conducted a mixed-effects analysis using a random intercept for the team, recognizing that offensive tendencies might differ by franchise. By examining correlation matrices and calculating VIF, we confirmed that collinearity was not problematic. Our modeling strategies allowed us to gauge the impact of within-team variability on daily outcomes while controlling for mean level measures and team-specific effects, thereby offering insights into whether a day's diversity in swing mechanics influences day-to-day offensive performance

Results

Individual-Level Findings (Univariate)

Model 1: $\text{xwOBA} \sim \text{mean_swing_length} + \text{sd_swing_length}$

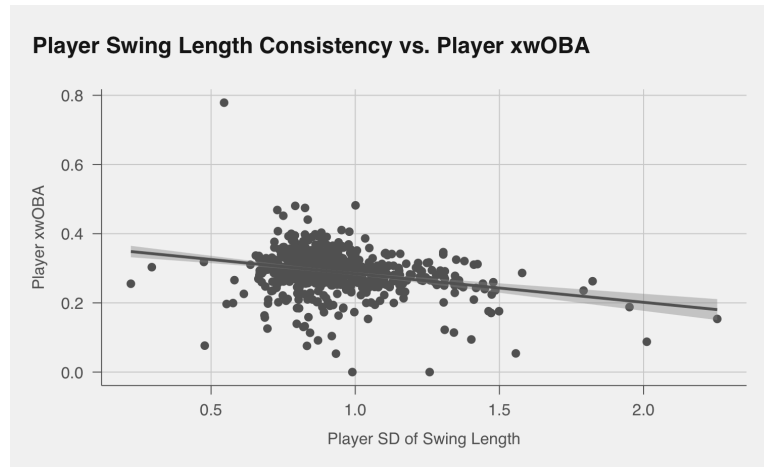


Figure 1. Player Swing Length Consistency vs Player xwOBA

The first model explored how a player's swing length consistency (measured by the standard deviation of swing length) and average swing length relate to offensive performance, as represented by xwOBA. The results highlight a significant negative relationship between swing length variability and xwOBA, suggesting that players who maintain more consistent swing lengths tend to perform better offensively. This model underscores the importance of mechanical consistency in driving offensive success over the course of a season. The negative coefficient for swing length variability could indicate that higher variability likely introduces inefficiencies or inaccuracies in swing mechanics, ultimately hindering performance.

Model 2: $\text{xwOBA} \sim \text{mean_bat_speed} + \text{sd_bat_speed}$

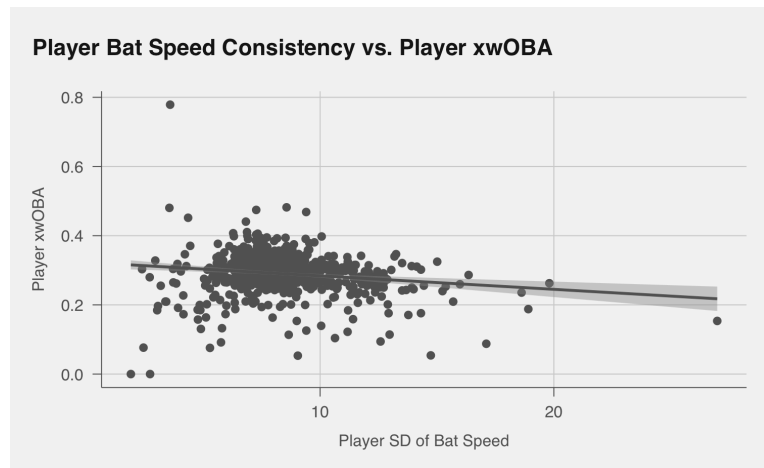


Figure 2. Player Bat Speed Consistency vs. Player xwOBA

In this model, we examined how variability and average bat speed affect xwOBA. Average bat speed is strongly and positively associated with offensive performance, indicating that players with faster bat speeds are generally more effective at producing higher xwOBA values. However, variability in bat speed does not have a statistically significant impact on xwOBA, suggesting that inconsistency in bat speed alone does not meaningfully detract from a player's performance when average bat speed is already accounted for. This result highlights a clear distinction between the effects of variability in swing length versus bat speed—variability in swing length appears more detrimental to performance than variability in bat speed.

Individual-Level Findings (Multivariate)

Model 3: $\text{xwOBA} \sim \text{mean_swing_length} + \text{mean_bat_speed} + \text{sd_swing_length} + \text{sd_bat_speed}$

This model includes both average and variability measures for swing length and bat speed, providing the most detailed analysis of the relationships between these mechanics and xwOBA for a player's season-level data. Both swing length variability and bat speed variability are significantly associated with performance, though their effects differ: higher swing length variability negatively impacts performance, while higher bat speed variability appears positively associated with xwOBA. This nuanced result suggests that, while consistency in swing length is crucial for success, there may be scenarios where some variability in bat speed is advantageous—possibly reflecting the ability to adapt bat speed to different pitch types. Average bat speed remains the strongest predictor of performance, further underscoring its importance.

Model 4: $\text{xwOBA} \sim \text{mean_swing_length} + \text{mean_bat_speed} + \text{mechanical_variability}$

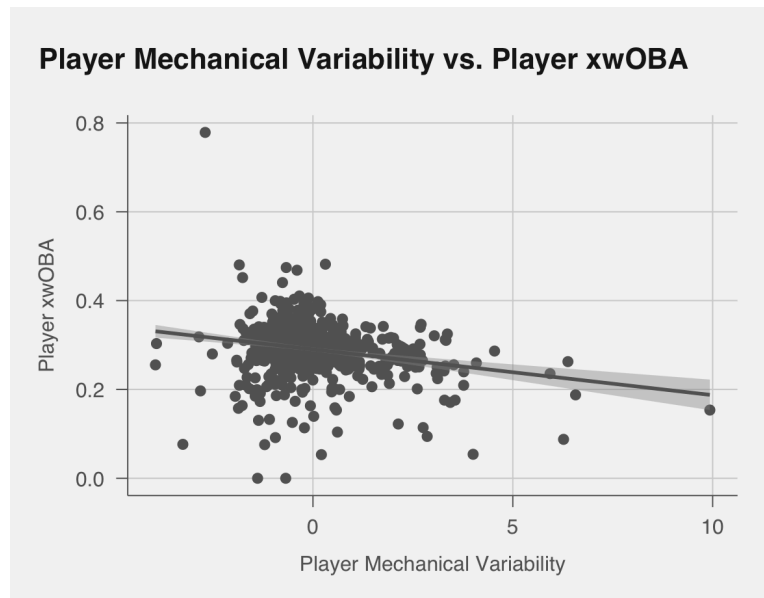


Figure 3. Player Mechanical Variability vs. Player xwOBA

This model introduces the “mechanical variability” component derived from PCA, which combines swing length and bat speed variability into a single metric. The negative relationship between mechanical variability and xwOBA supports the broader idea that inconsistency in mechanics hinders offensive output. Interestingly, when both average swing length and bat speed are included in the model alongside mechanical variability, only average bat speed remains strongly predictive of performance, while swing length loses its significance. This suggests that mechanical variability offers a more holistic measure of how consistent mechanics impact performance. The model’s findings emphasize that players with steadier mechanics across both dimensions—swing length and bat speed—tend to generate better offensive results.

Game-Level Findings (Univariate)

Model 5: $\text{game_xwOBA} \sim \text{game_mean_median_swing_length} + \text{game_sd_median_swing_length}$

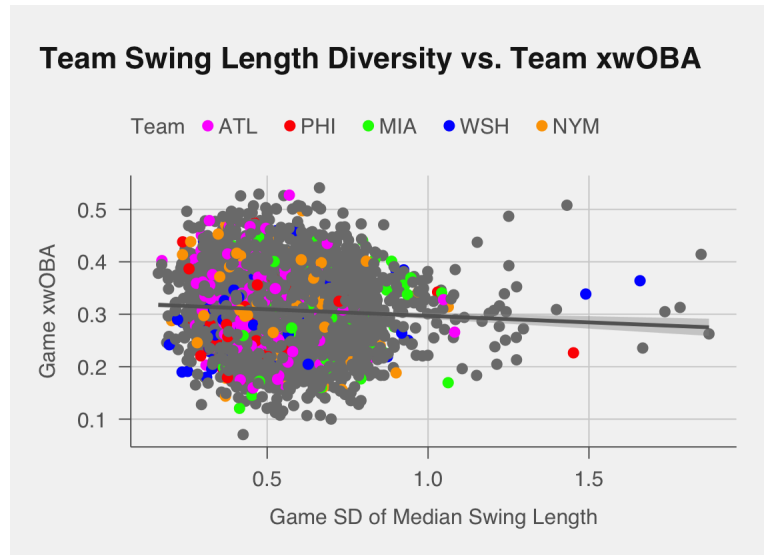


Figure 4. Team Swing Length Diversity vs. Team xwOBA, where each data point represents a game. NL East teams are colored for reference.

In this model, we explored the impact of team-level swing length metrics—both the mean of median swing lengths and the variability (SD of median swing lengths)—on game-level offensive performance, represented by xwOBA. The findings indicate that higher average swing lengths at the team level are positively associated with offensive performance, suggesting that teams with longer swing lengths tend to generate better outcomes. Conversely, greater variability in swing lengths negatively affects xwOBA, implying that teams with inconsistent swing lengths struggle to produce optimal offensive results. Together, these results emphasize the importance of both maintaining a longer swing length on average and reducing variability across players' swing lengths to maximize team performance.

Model 6: $\text{game_xwOBA} \sim \text{game_mean_median_bat_speed} + \text{game_sd_median_bat_speed}$

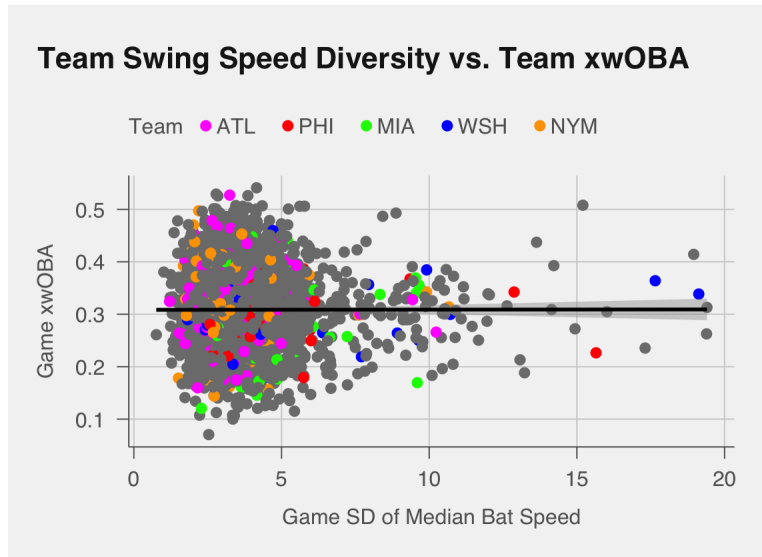


Figure 5. Team Swing Speed Diversity vs. Team xwOBA, where each data point represents a game. NL East teams are colored for reference.

This model analyzed the relationship between team-level bat speed metrics and xwOBA. Both the mean of median bat speeds and the variability (SD of median bat speeds) are positively associated with xwOBA. These findings suggest that teams with higher average bat speeds tend to perform better offensively, highlighting the critical role of bat speed in generating productive at-bats. Interestingly, variability in bat speed also positively correlates with offensive performance, which might indicate that some degree of lineup adaptability or dynamic adjustment in bat speed across players contributes to better outcomes. This result contrasts with the negative effect of swing length variability seen in the first model, suggesting that variability may play different roles depending on the mechanical characteristic in question.

Model 7: $\text{game_xwOBA} \sim \text{game_sd_median_bat_speed} + \text{game_sd_median_swing_length} + \text{game_mean_median_bat_speed} + \text{game_mean_median_swing_length}$

In this model, we included both swing length and bat speed metrics—means and variability—to assess their combined impact on game-level xwOBA. The results show that average bat speed remains a strong positive predictor of offensive performance, while variability in swing length continues to have a significant negative impact. Additionally, variability in bat speed retains its positive association with xwOBA, while average swing length shows a weaker and less consistent relationship. These findings reinforce the idea that bat speed, particularly higher

averages, is a critical driver of performance, while swing length consistency remains vital. The contrasting roles of variability between bat speed (positive) and swing length (negative) suggest nuanced mechanics that teams should consider when optimizing offensive strategies.

Model 8: $\text{game_xwOBA} \sim \text{game_sd_median_bat_speed} + \text{game_sd_median_swing_length} + \text{game_mean_median_bat_speed} + \text{game_mean_median_swing_length} + (1 \mid \text{batter_team})$

This final model incorporated random effects for teams to account for unobserved differences between teams that might influence game-level offensive performance. Similar to the previous model, average bat speed remains a strong and positive predictor, while swing length variability continues to show a negative relationship with xwOBA. Variability in bat speed also retains its positive association, even after accounting for team-level effects. The inclusion of team-specific random effects helps to mitigate that certain teams may possess inherent characteristics, such as coaching strategies or player composition, that influence their overall offensive output. This model highlights the consistency of the relationships found across different modeling approaches, while also acknowledging the role of team-level factors in shaping game outcomes.

Conclusion

Takeaway #1: Consistency in swing length is linked to better performance for a player over a season.

Across our individual-level analyses, a clear pattern emerges: players who maintain a more stable swing length (i.e., lower standard deviation in swing length) tend to produce higher xwOBA. On a game-by-game basis, teams that collectively show less variability in swing length appear more likely to generate stronger offensive outcomes. These findings suggest that consistency in swing length is beneficial at both the individual and group levels.

Takeaway #2: Variability in bat speed can be advantageous

In contrast to swing length, where increased variability is consistently detrimental, some degree of fluctuation in bat speed emerges as potentially beneficial. At the individual level, bat speed variability does not significantly hurt performance when controlling for average bat speed—and in certain models, it even shows a positive association with xwOBA. At the game level, teams displaying a wider range of bat speeds can still put up strong offensive performances. This

dynamic suggests that a lineup's ability to adjust to different pitcher types might be enhanced by more variable bat speeds amongst its players.

Takeaway #3: A Holistic View of Swing Mechanics Matters

Our regression models using a combined “mechanical variability” component (derived via PCA) reveal that stability across both bat speed and swing length is crucial. While bat speed remains a strong predictor on its own, incorporating an overall measure of mechanical variability further refines our understanding of how a hitter's mechanics translate to performance. Teams would do well to monitor both dimensions—bat speed and swing length—to optimize their hitters' performance.

Discussion

Implications on Player Development

From a developmental standpoint, these findings underscore the importance of working with hitters on achieving and maintaining a consistent swing length, while simultaneously cultivating sufficient bat speed. Drills, individualized training sessions, and biomechanics evaluations can help players refine their swing mechanics and reduce extraneous movements that might introduce excessive swing length variability. However, the positive correlations surrounding bat speed variability suggest that it could be advantageous for hitters to develop an adaptable swing approach—one that can increase or decrease bat speed as needed. Hitting coaches might consider incorporating situational training to encourage players to make purposeful adjustments in bat speed while keeping their swing length relatively stable. It is worth noting that our results do not show causality, but merely a suggestion of a possible link. It is possible that the best hitters have reached optimal swing styles that make them the best hitters, resulting in a sort of circular logic where it is hard to uncover cause and effect. However, our study shows a need for further analysis on player development strategies focusing on communicating a consistency, rather than adjustment of swing length throughout a season.

Implications on Lineup Strategy

For teams, the results point to intriguing possibilities in daily lineup construction and overall roster composition. Managers might look to ensure that the collective swing length variability

remains controlled, and on days when certain matchups or pitch types are anticipated, lineups could be optimized to capitalize on hitters who are not only capable of generating above-average bat speeds but also optimized to have a wide variety of bat speeds to face a variety of pitchers in every game.

Limitations

Although our analyses reveal suggestive correlations between mechanical consistency and offensive performance, the results should be interpreted with caution. First, we cannot claim causality; more consistent hitters might simply be better hitters overall, and external factors such as pitch mix, ballpark effects, or pitcher quality may influence performance in ways not captured by our models. Second, our game-level measures are based on aggregated statistics, which can mask individual-level nuances such as platoon advantages, in-game adjustments, or injury status. Future research should investigate these limitations, ideally through controlled studies or advanced in-game tracking to isolate the precise mechanical changes that drive offensive output.

References

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