

# Beyond Expected Goals: A Probabilistic Framework for Shot Occurrences in Soccer

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# What Are Expected Goals (xG)?

- Expected Goals (xG) is a metric that estimates the probability that a shot is scored
- Depends on factors like distance from goal, angle to goal, shot type, and player positions
- Estimated by XGBoost models trained on historical shot data
- Has many practical uses
  - A measure of individual shot quality
  - Provides many individual-player metrics
  - A comparison of team performance in a match
  - A measure of team performance over a season

# Limitations of xG

- Models are only trained on **observed** shots, inducing significant selection bias
  - Skilled attackers who take more shots are over-represented
  - Threatening attacks with no recorded shots are omitted
- Aggregating xG additively biases performance estimates
  - Rebounded chances are double-counted
  - Players only receive credit when a shot occurs

# Visual Example 1

*Video: Player has clear chance but passes instead of shooting*

## Visual Example 2

*Video: Multiple shots from same attacking sequence w/ xG that add over 1*

# Our Target Metric: xG+

- A more complete picture of goal expectancy
  - Accounts for high-threat attacks with no shots
  - Does not double-count rebound chances
- At each frame  $t$ , we calculate xG+ as the probability a goal is scored

$$\begin{aligned}\text{xG+}_t &= \mathbb{P}_t(\text{goal scored}) \\ &= \mathbb{P}_t(\text{goal scored} \mid \text{shot taken}) \cdot \mathbb{P}_t(\text{shot taken}) \\ &= \text{xG}_t \cdot \text{xShot}_t\end{aligned}$$

- Then over a possession with  $n$  frames, xG+ is defined as

$$\text{xG+}_{\text{poss}} = 1 - \prod_{t=1}^n [1 - \mathbb{P}_t(\text{goal scored})]$$

# Estimating xShot

- xShot: the probability that a shot occurs in the next second
- Build a model to estimate xShot based on features from tracking data
- Also build our own version of xG model using the same features on observed shots

- Remove games where no shots are recorded
- Only keep frames where the ball is in play and a team has clear possession
- Linearly interpolate ball positions to fill in missing frames
- attack: Index of the attack the current frame is on (0 if it is not on an attack)
  - Start with the attacking team gaining possession in their attacking third
  - End with the defending team regaining possession or the ball is out of their attacking third
  - Only keep frames with  $\text{attack} > 0$



- Rotate the coordinates  $180^\circ$  around the center point for frames where the team attacks from right to left to unify the attacking directions and make all  $x$ -coordinates positive
- Use a polar coordinate system centered on the goal for the ball
  - $r_{ball}$  and  $\theta_{ball}$  represent the distance and angle of the ball from the goal
  - Keep the  $z$ -coordinate and compute the speed of the ball
- Use a polar coordinate system centered on the ball for each player
  - Choose the 5 closest offense teammates and non-GK defenders to the ball as features
  - Keep goalkeeper positions as a separate feature

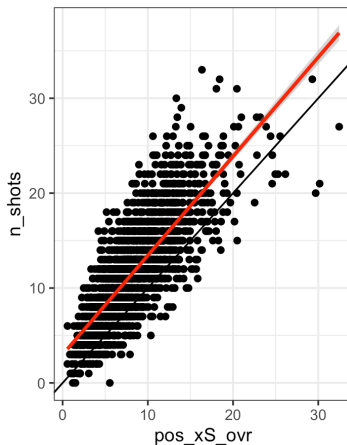
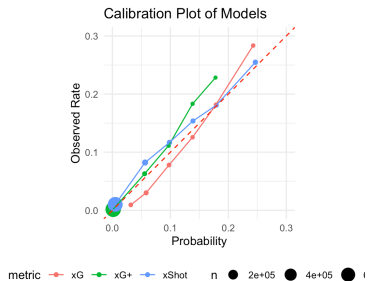
- `openGoal`: Percentage of the goal that is open from the ball's position
  - Simplify every defender as a circle with a radius of 0.75 m
  - Compute the two tangent lines from the ball to every defender in front of the ball their intersection points with the goal line
  - Calculate the length of the open goal as the length of goal not covered by segments formed by the intersection points

# Model Specifications

- Trained on all tracking data of 2022-2025 Premier League seasons
- Use a 5-fold cross-validation to evaluate both xG and xShot XGBoost models
- Choose log loss as the evaluation metric

- Key findings
- Statistical significance
- Practical implications

# Results



# Conclusions

- Summary of main points
- Future work
- Questions and discussion

# Thank You

Questions?