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

Tianshu Feng, Jonathan Pipping, and Paul Sabin

University of Pennsylvania

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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
- Often used to measure the quality of a chance
- Aggregated over a match or season to measure team performance



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 across a match double-counts rebound chances



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
- ullet At each frame t, we calculate xG+ as the probability a goal is scored

$$\begin{aligned} \mathbf{x} \mathbf{G} +_t &= \mathbb{P}_t(\mathsf{goal\ scored}) \\ &= \mathbb{P}_t(\mathsf{goal\ scored}\mid \mathsf{shot\ taken}) \cdot \mathbb{P}_t(\mathsf{shot\ taken}) \\ &= \mathbf{x} \mathbf{G}_t \cdot \mathbf{x} \mathsf{Shot}_t \end{aligned}$$

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

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



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



Data Processing

- 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 attack > 0



Data Processing

- ullet Rotate the coordinates 180° 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
 - \bullet $\;r_{ball}$ and θ_{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



Data Processing

- 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

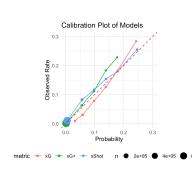


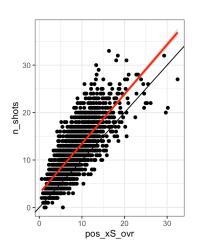
Results

- Key findings
- Statistical significance
- Practical implications



Results







Conclusions

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



Thank You

Questions?

