

# Evaluation of Off-the-Ball Actions in Soccer

Lucas Wu<sup>1,2</sup> and Dr. Tim Swartz<sup>1</sup>

<sup>1</sup> Simon Fraser University

<sup>2</sup> Zelus Analytics

# Motivation

- Individual players have possession of the ball for less than two minutes (Link and Hoernig, 2017).

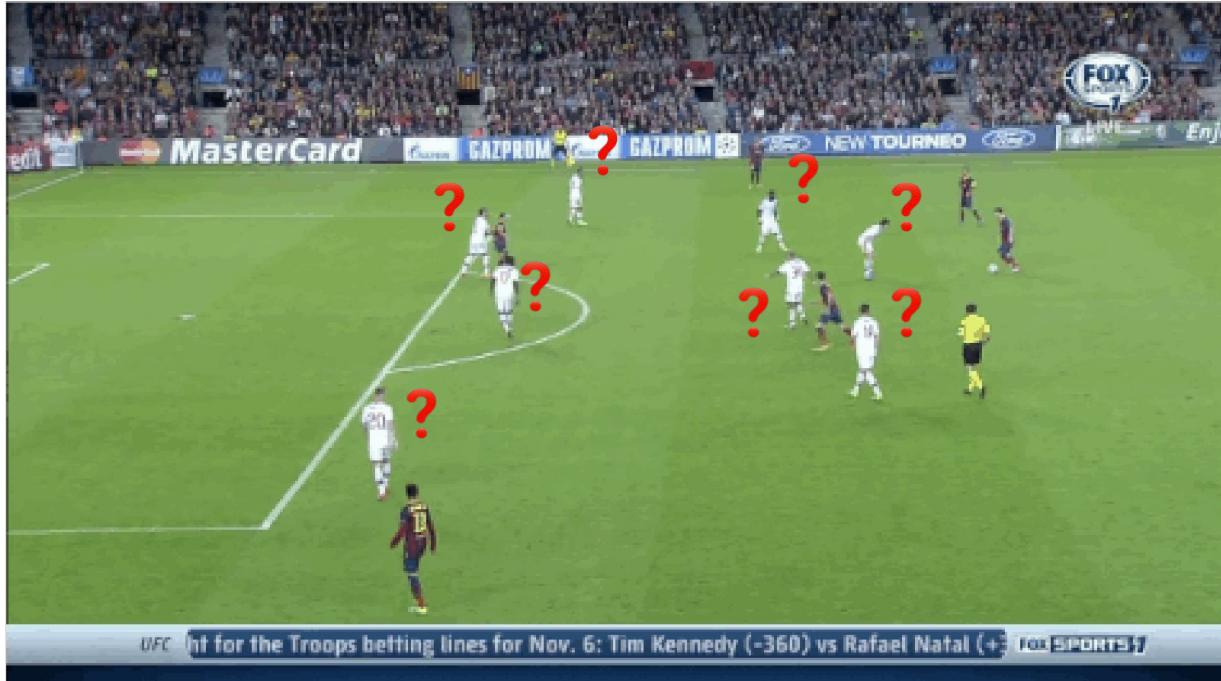
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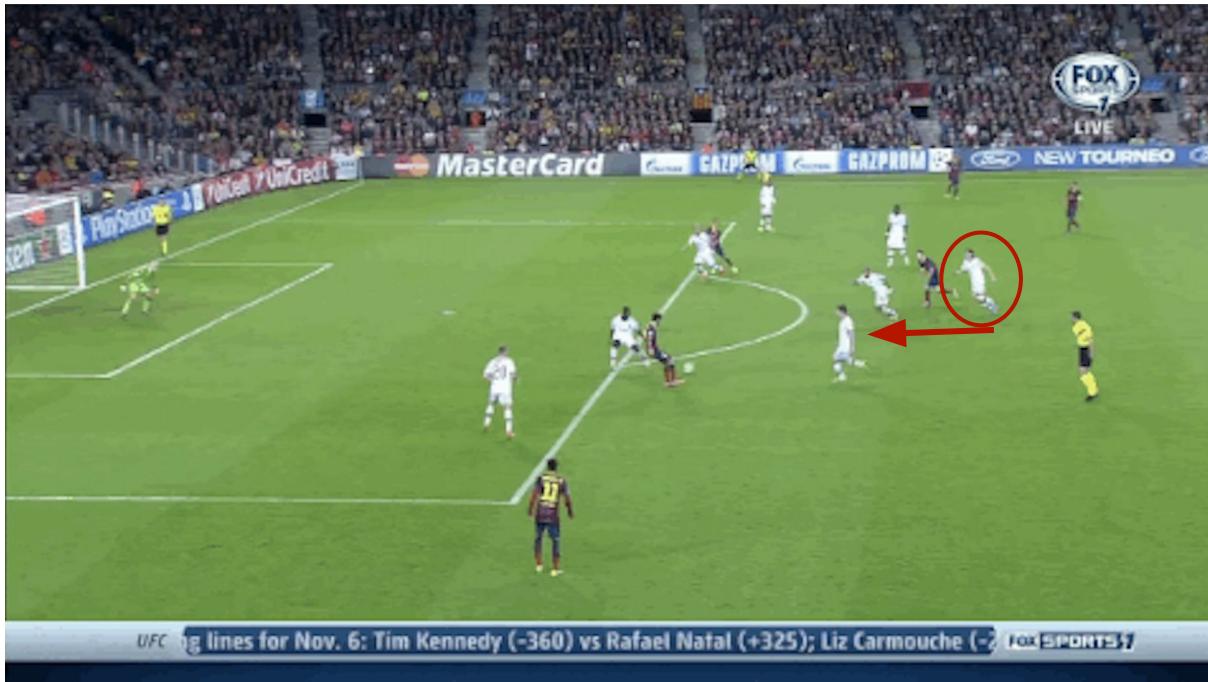
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UFC for Nov. 6: Tim Kennedy (-360) vs Rafael Natal (+325); Liz Carmouche (-210) vs FOX SPORTS 1

# Rationale & Objective

- Evaluate how defensive players move in a given situation (defensive anticipation)
- Recognize similar match situations to understand what an average defender would do

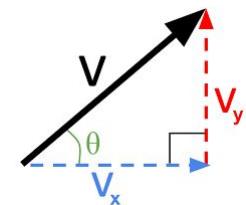
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- Recognize similar match situations to understand what an average defender would do
- Prediction of velocity given a set of spatial-temporal covariates/features

$$V = f(X)$$



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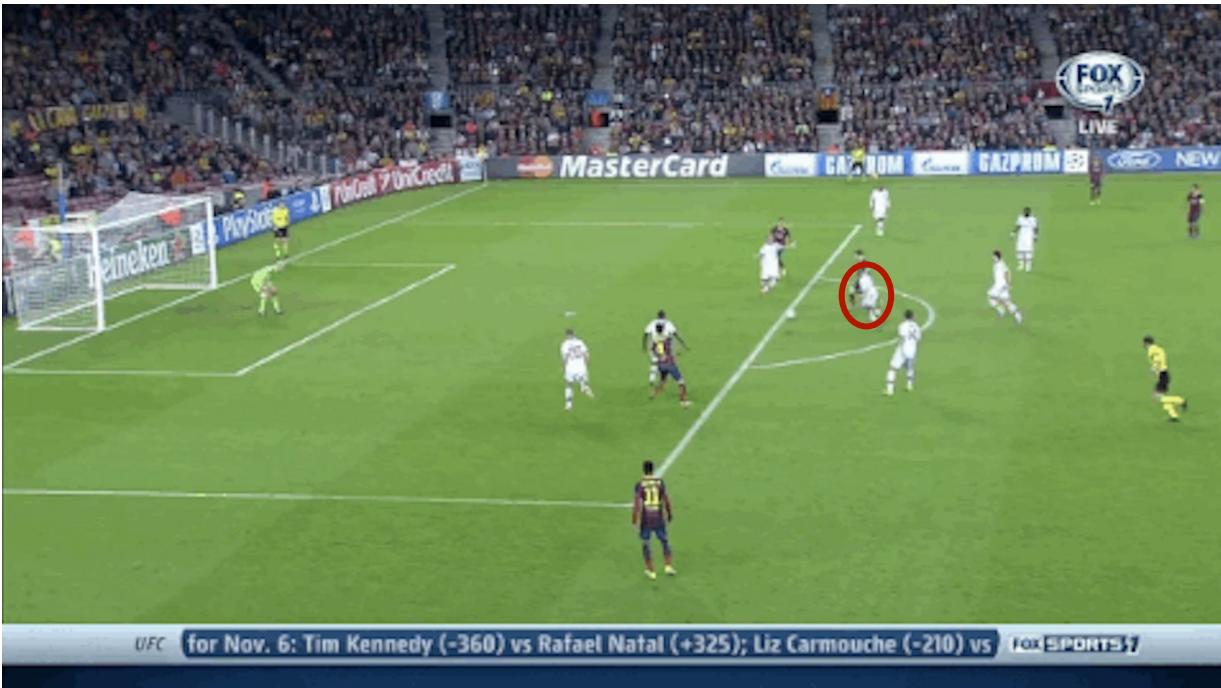


# Covariates/Features

$$V = f( )$$



- $f()$  could be a highly complex function
- Hand-crafted covariates/features based on soccer knowledge
  - Player of interest (fixd a given defender), ball, teammates, opponents, game context



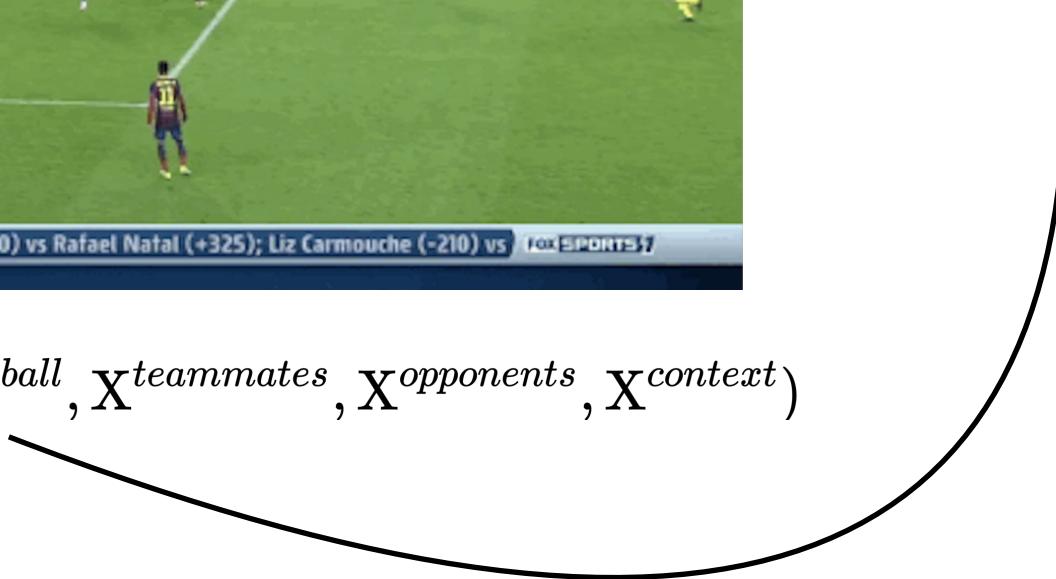
- x,y coordinates
- player velocity half seconds before
- distance of the player to the ball
- distance of the player to offensive goal
- angle of the player to offensive goal

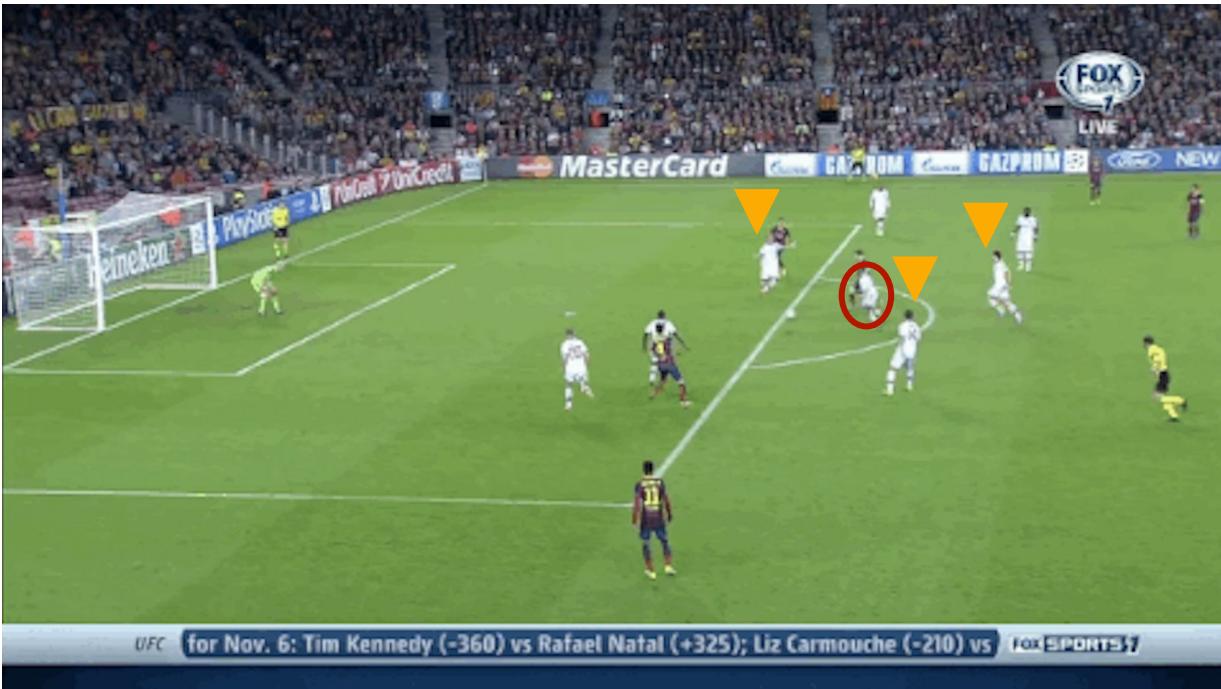
$$V = f(X^{player}, X^{ball}, X^{teammates}, X^{opponents}, X^{context})$$



- x,y coordinates of the ball
- ball speed

$$V = f(X^{player}, X^{ball}, X^{teammates}, X^{opponents}, X^{context})$$





### - 3 nearest teammates

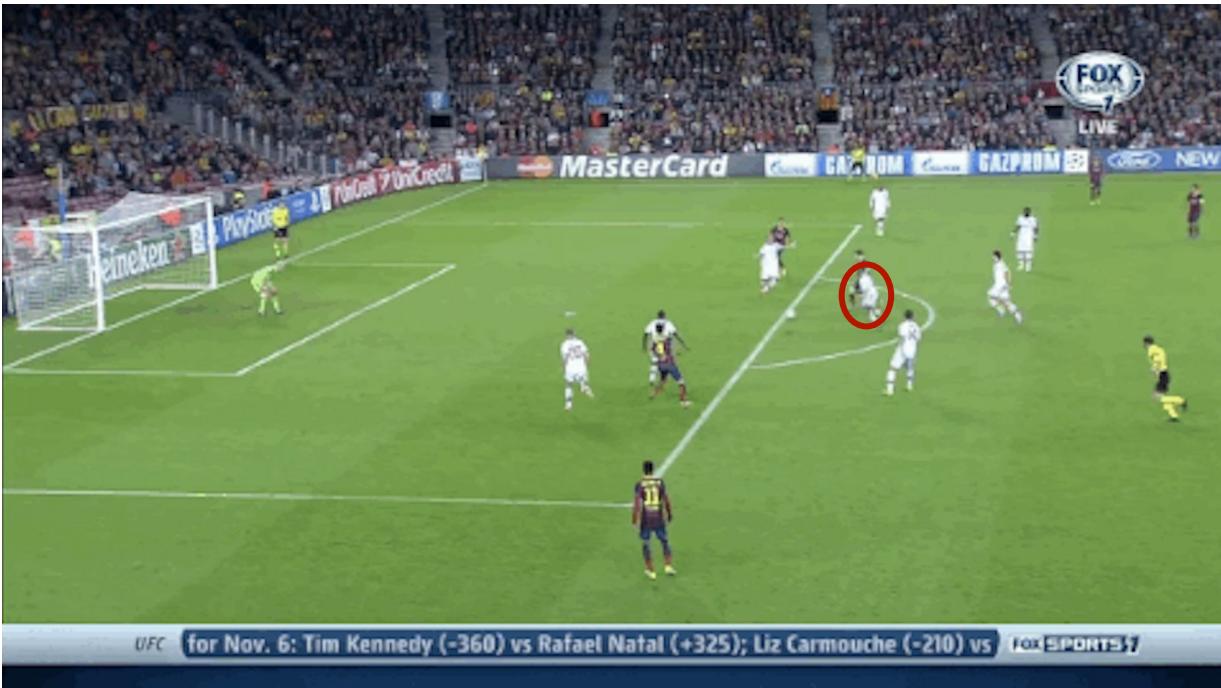
- x,y coordinates
- velocity of teammates
- distance of the player to teammate
- distance of the ball to teammate
- relative angle of the teammate to the player of interest

$$V = f(X^{player}, X^{ball}, X^{teammates}, X^{opponents}, X^{context})$$



- 3 nearest opponents
  - x,y coordinates
  - velocity of opponent
  - distance of the player to opponent
  - distance of the ball to opponent
  - relative angle of the opponent to the player of interest
  - $xT$  at the current location
- x,y coordinates of the goalkeeper
- distance of the player to goalkeeper

$$V = f(X^{player}, X^{ball}, X^{teammates}, X^{opponents}, X^{context})$$



- home or road team
- seconds remaining in the half of the game
- seconds remaining in the full game
- indicator for the player on offensive or defensive side of the field

$$V = f(X^{player}, X^{ball}, X^{teammates}, X^{opponents}, X^{context})$$

# Data

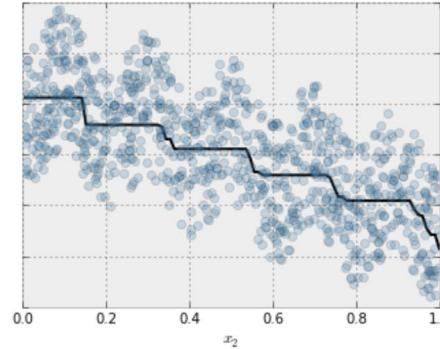
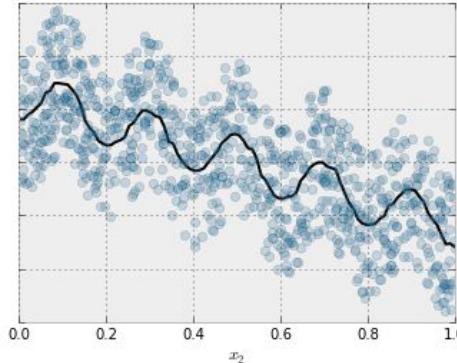
- Player tracking data Chinese Super League in 2019 season (~200 games)
- 61 covariates/features
  - $\mathbf{X}^{player}, \mathbf{X}^{ball}, \mathbf{X}^{teammates}, \mathbf{X}^{opponents}, \mathbf{X}^{context}$
- 10 times per second -> downsample to every second
- Focus on the case when opponent has maintained possession

# Model

- Gradient boosting model
  - Works well on tabular data empirically
    - 17/29 Kaggle winning solutions in 2015 used XGBoost (Chen & Guestrin, 2016)

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  - Works well on tabular data empirically
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  - Faster to train and tune
  - Enforce monotonicity constraints on features

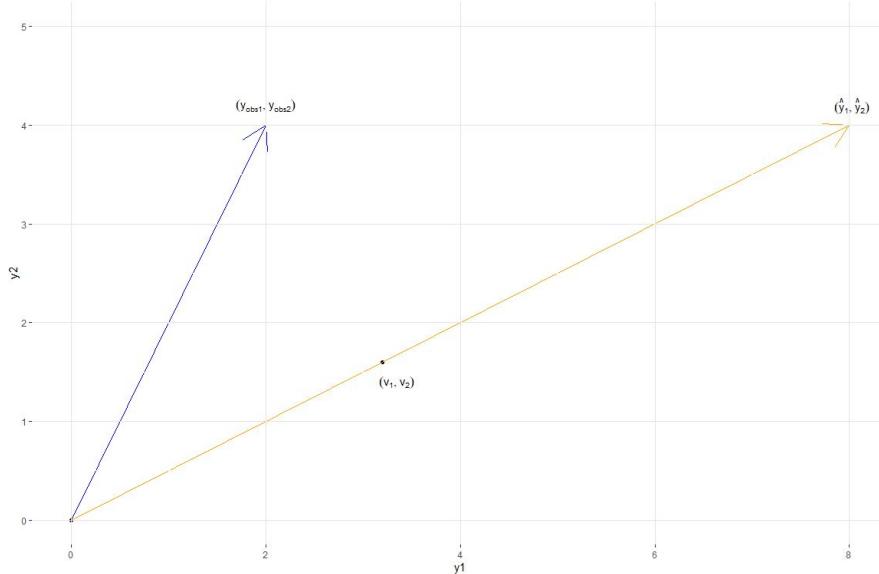


# Defensive anticipation metric

For a given  
player

$$p = \begin{cases} \left( \sqrt{v_1^2 + v_2^2} - \sqrt{\hat{y}_1^2 + \hat{y}_2^2} \right) / \sqrt{\hat{y}_1^2 + \hat{y}_2^2} & v_1 \hat{y}_1 \geq 0 \\ \left( -\sqrt{v_1^2 + v_2^2} - \sqrt{\hat{y}_1^2 + \hat{y}_2^2} \right) / \sqrt{\hat{y}_1^2 + \hat{y}_2^2} & v_1 \hat{y}_1 < 0 \end{cases}$$

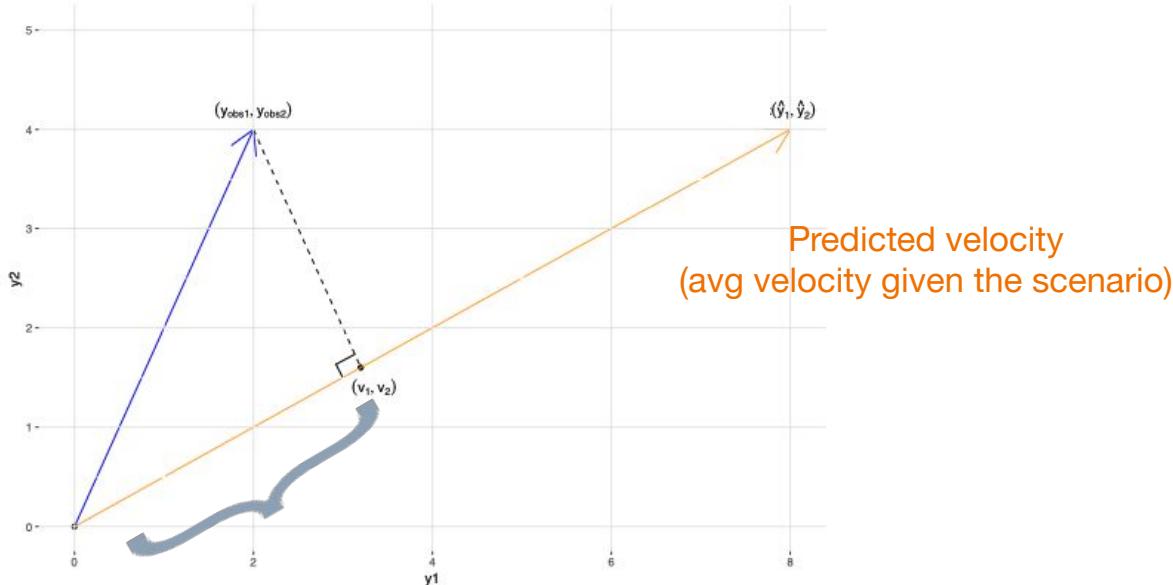
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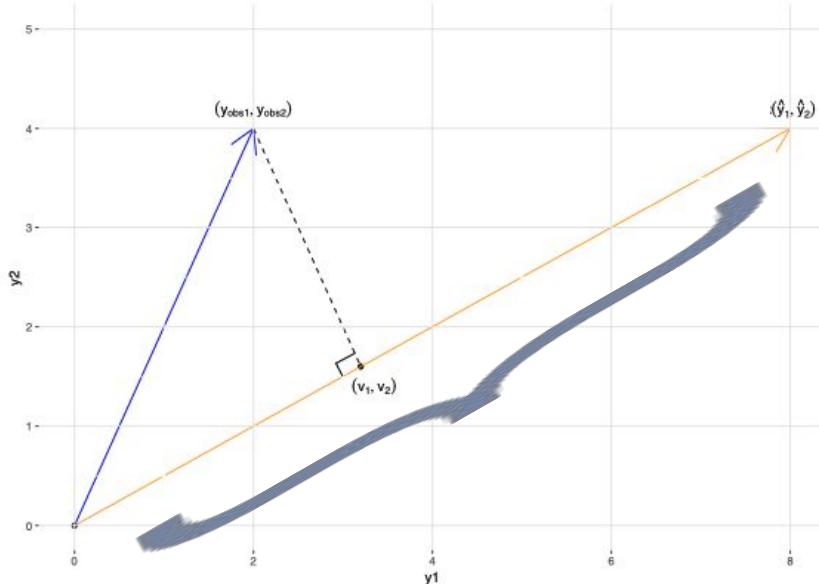
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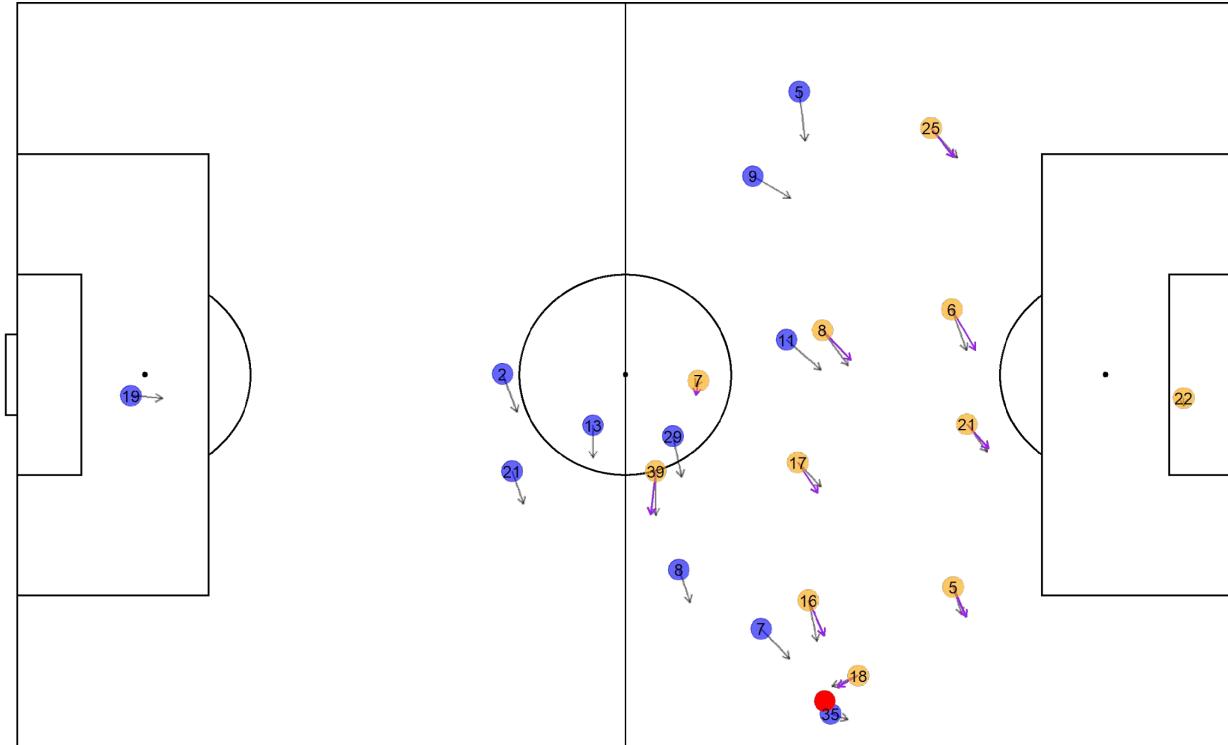
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# Results

## Defensive anticipation metric for all defenders

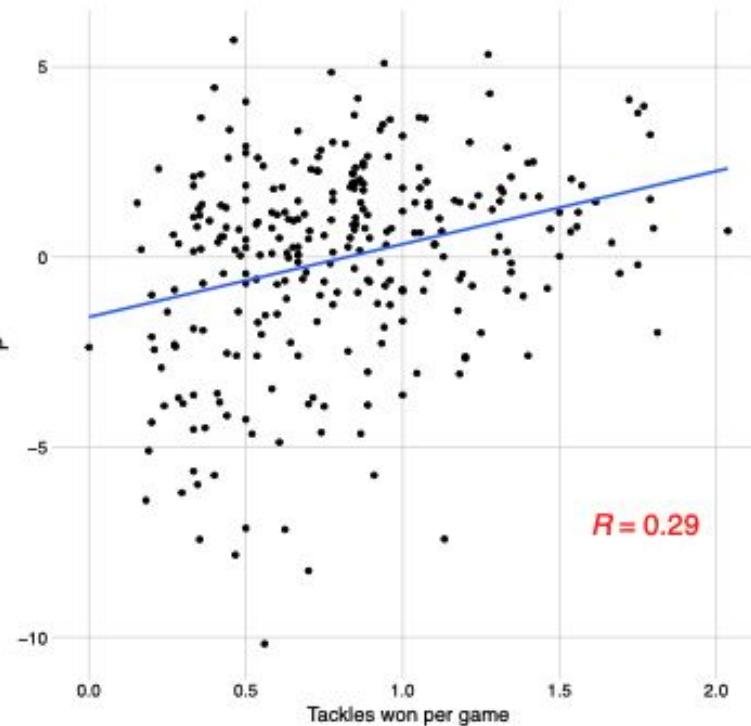
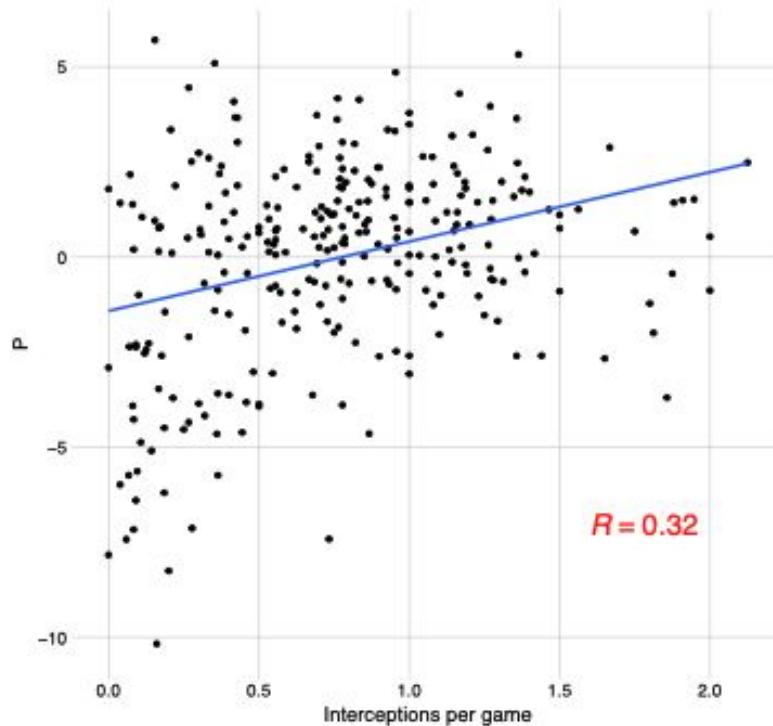
Blue: Off team  
Yellow: Def team

Purple: Predicted velocity  
Grey: Observed velocity



# Results

## Correlation of defensive anticipation vs interceptions/tackles



# Discussion & Future work

- Spilt data into odd/even weeks and out-of-sample player rankings are consistent
- Combine the predicted velocity with pitch control models
- Condition on soccer tactics and styles of play