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# Quantifying Pitch Control

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# Quantifying Pitch Control

William Spearman

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# Introduction

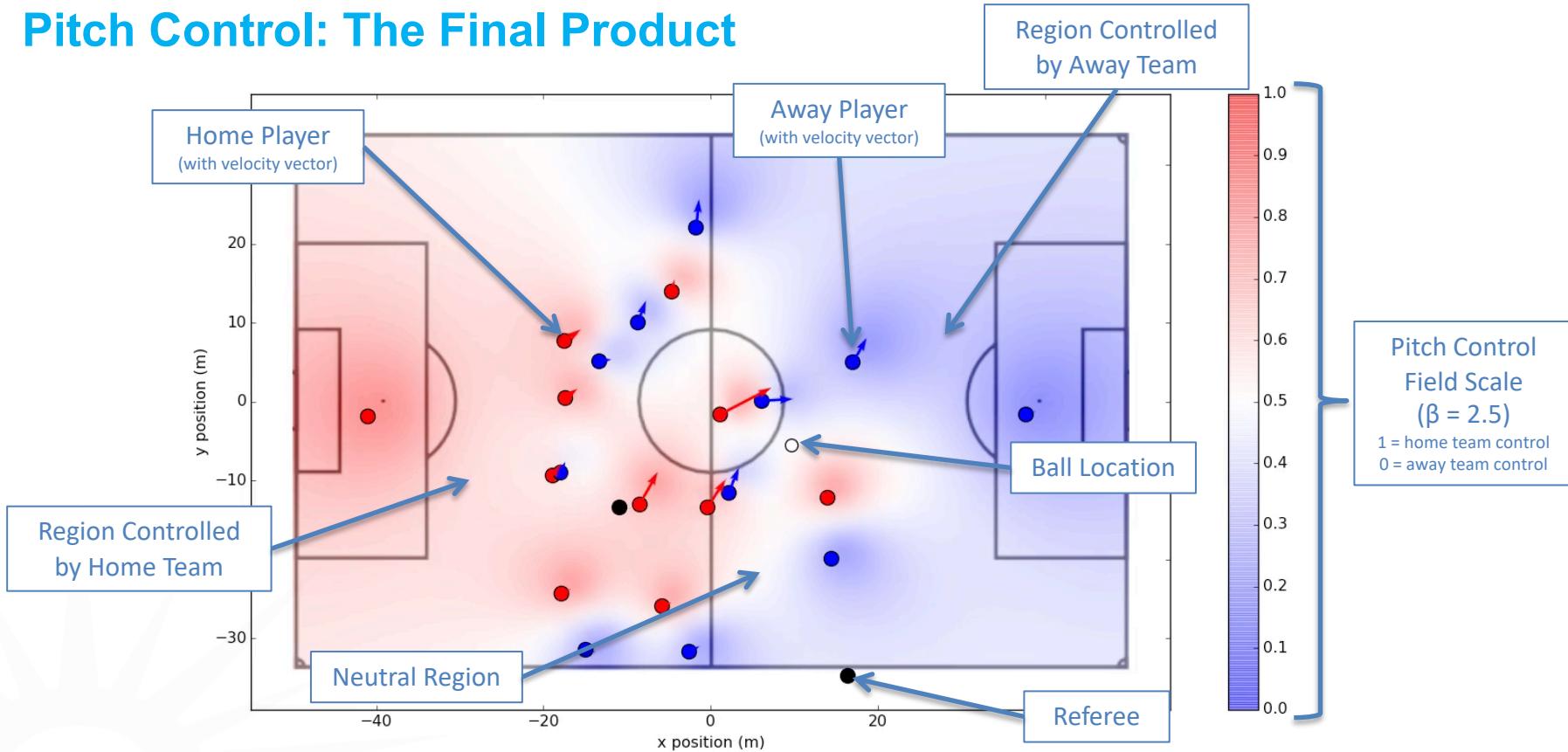
## Dr. William Spearman

- Ph.D. in High Energy Particle Physics from Harvard University
  - Studied the Higgs Boson
- Works as a data scientist for Hudl
  - Not a football expert!

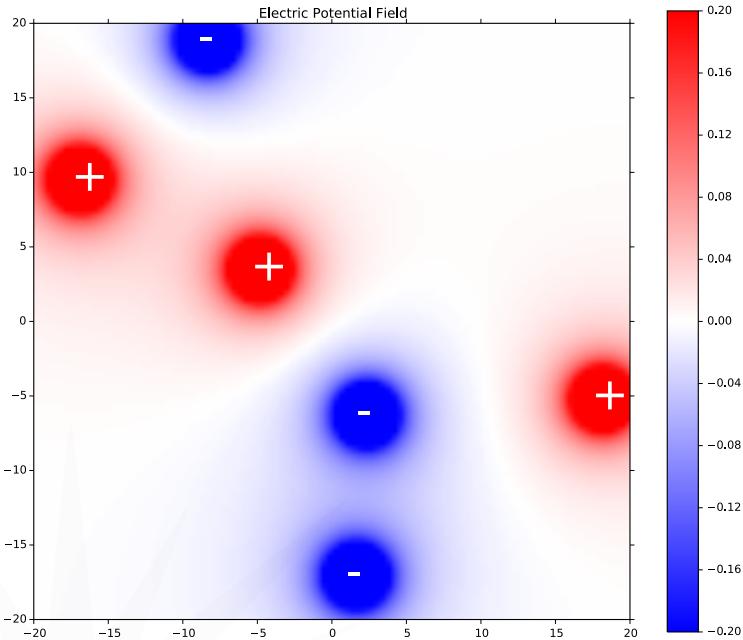


Hudl + Replay Analysis + Sportstec

# Pitch Control: The Final Product



# Motivation



## Physics

The **electric potential field** quantifies the way charged **particles** exert a force on a test charge in **space**.

## Football

We propose the development of a **pitch control field** that can be used to quantify the way football **players** control regions on the **pitch**.

## What is Pitch Control?

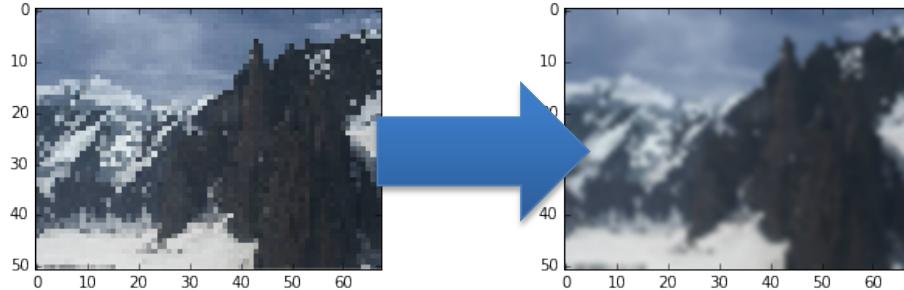
### Definition

We define *pitch control field (PCF)* for location,  $x$ , as the probability that the home team will end up with possession of the ball if it were at location,  $x$ . The PCF predicts the “next possessor”.

### Get there first with the most men

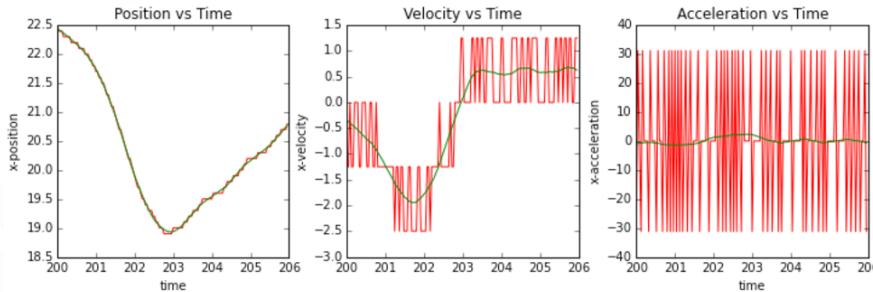
The PCF will probably depend on **the time** it takes each player to reach location,  $x$  and this time will depend on the location and velocity of each player.

# Using Tracking Data



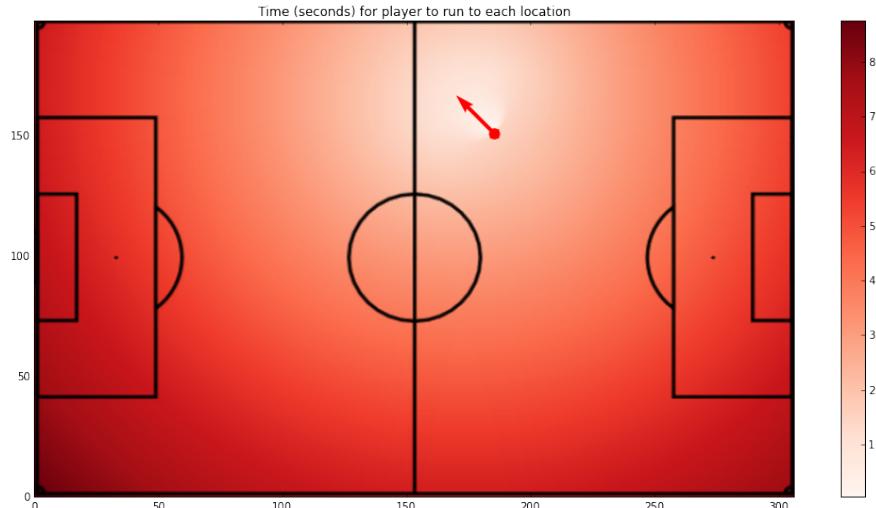
## TRACAB Data

- Player/ball positions at 25 fps
- We smooth it with an S-G filter



## Calculating Times Using

- Player position
- Player velocity
- Player acceleration
- Maximum player speed



# Choosing a Model

Label for the  $i$ th player (1 for Home team and -1 for away team)

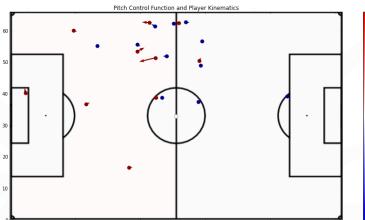
$\beta$  is a parameter which indicates how much to weight being the first to the ball (Range = 0 to  $\infty$ )

$$PCF(t_i, l_i) = \left[ \frac{\sum_i l_i t_i^{-\beta}}{\sum_i t_i^{-\beta}} + 1 \right] / 2$$

Time for the  $i$ th player to reach the ball

This part will be between -1 and 1

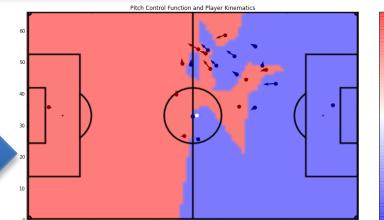
Use a naïve linear scaling to change output from -1 to 1  $\rightarrow$  0 to 1.



## Understanding the Parameter, $\beta$

$\beta = 0$ : PCF always 0.5

$\beta = \infty$ : PCF is 1 if closest player is on the home team, otherwise, it's 0.



# Fitting Strategy

## 1. Sync Opta and Tracab

- Identify when the ball is “in-play”

## 2. Calculate ball’s possessor

- Contested – both teams are near it
- None – no team is near the ball
- Home/Away – one team has uncontested control of the ball

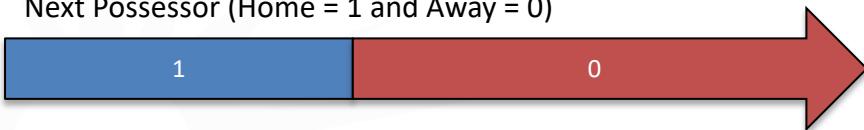
Current Possessor



## 3. Calculate ball’s *next* possessor

- This is done by looking forward in time to see which team gets the next uncontested possession.

Next Possessor (Home = 1 and Away = 0)



## 4. Choose Fit Frames

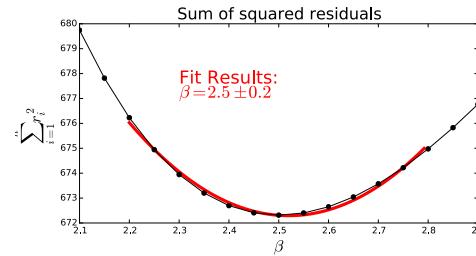
- We focus on frames where there *possession* is *contested* or *none*.
- We want there to be a clear *next possessor*.
- The next possessor is *truth* value for who gets the ball next

## 5. Calculate PCF at ball’s location.

- This gives us the model’s *prediction* for who gets the ball next.

## 6. Minimize sum of squared errors

- Calculate residual for each frame:
  - $r = x_{pcf} - x_{real}$
- Sum of the squared residuals for each frame,  $i$ :
  - $\Sigma r^2 = \sum_i (x_{pcf,i} - x_{real,i})^2$ .



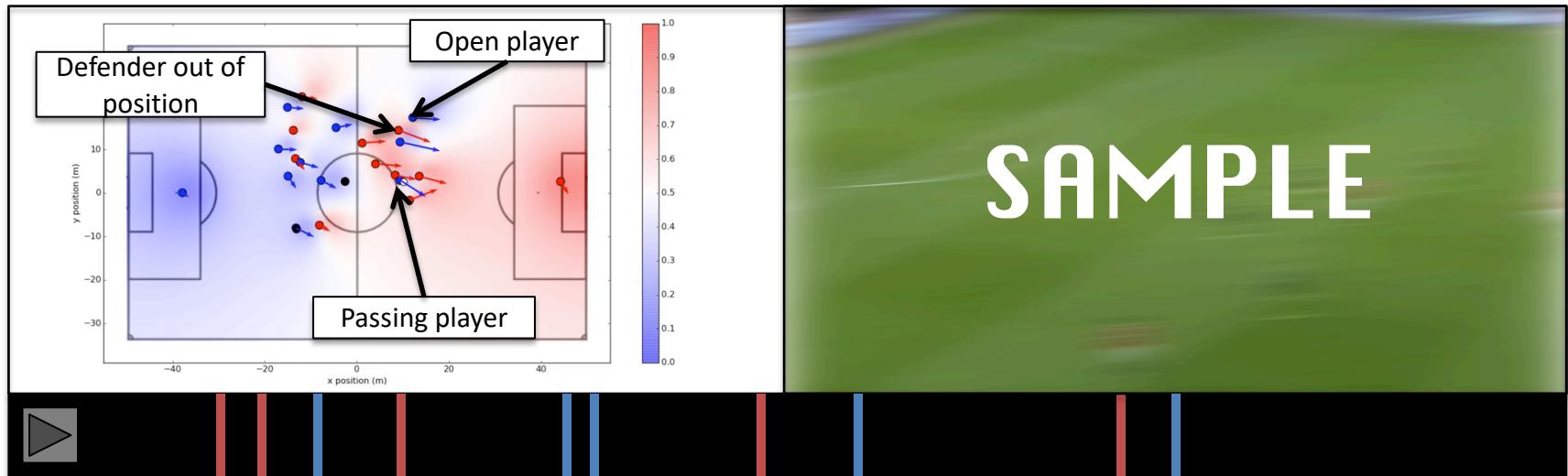
# Applications: A New Way To Watch Film

## Watch Alongside Film

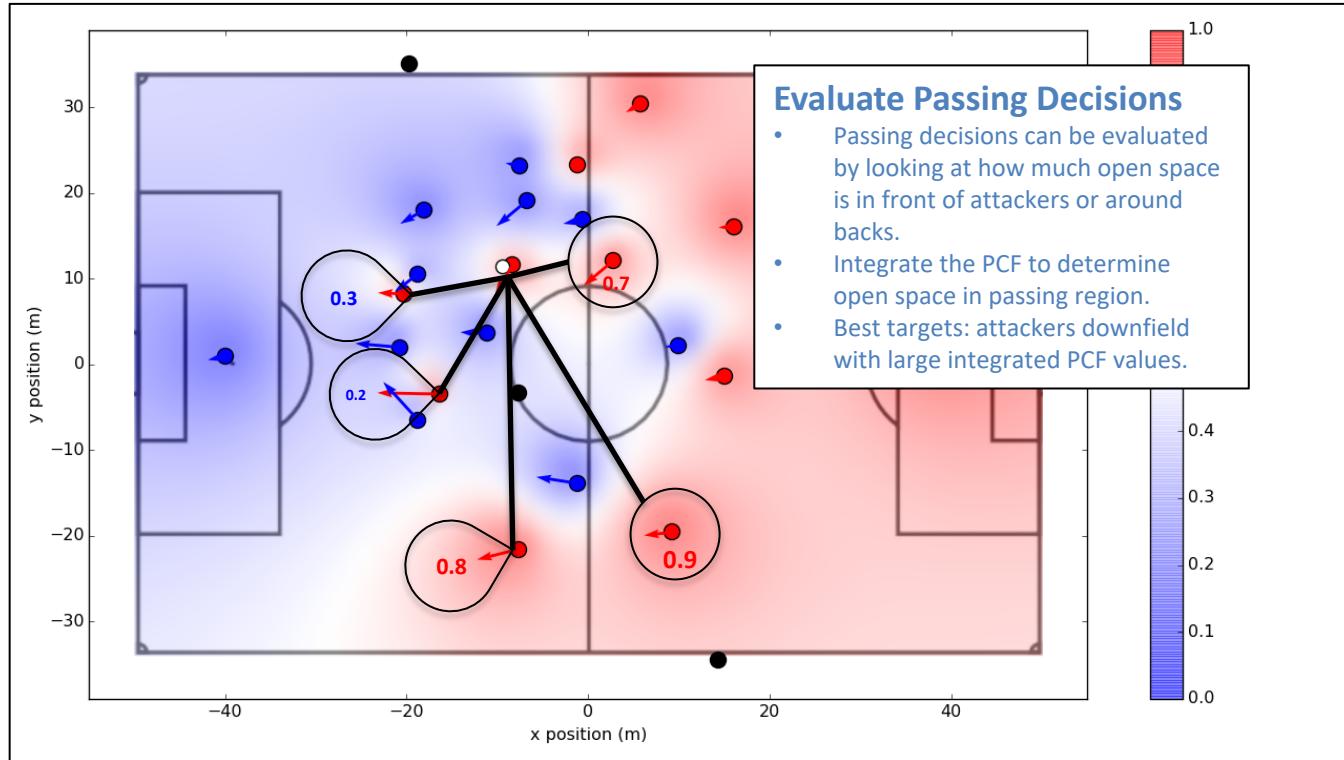
- Watch animated PCF along with video.
- PCF makes it possible to visualize space in-between units and space behind units.

## Identify

- Defensive players who are out of position (even if the mistake doesn't result in a goal)
- Missed offensive opportunities.



# Applications: A New Way To Watch Film



# Applications: A New Metric for Player Performance

## Identify Controlling Players

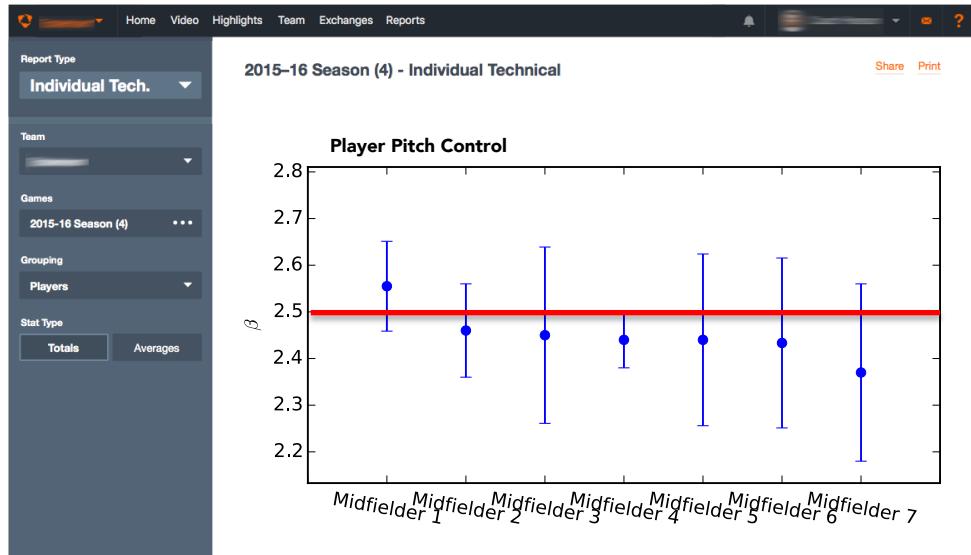
- Certain players are more capable of contesting for the football than others.
  - This should appear if we calculate their “player-specific” beta values.
  - In other words, how much does their presence improve their team’s chance of gaining possession of the ball above the average.

## How?

- This is done by fitting  $\beta_i$  separately for each player,  $i$ .

$$PCF(t_i, l_i) = \left[ \frac{\sum_i l_i t_i^{-\beta_i}}{\sum_i t_i^{-\beta_i}} + 1 \right] / 2$$

- Need more statistics to make inferences about specific player trends.
- $\beta$  only matters w.r.t. other players. No common sense interpretation.



## Results

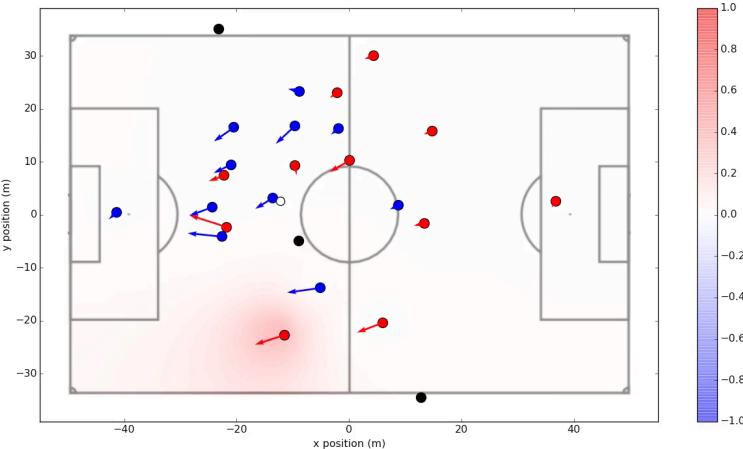
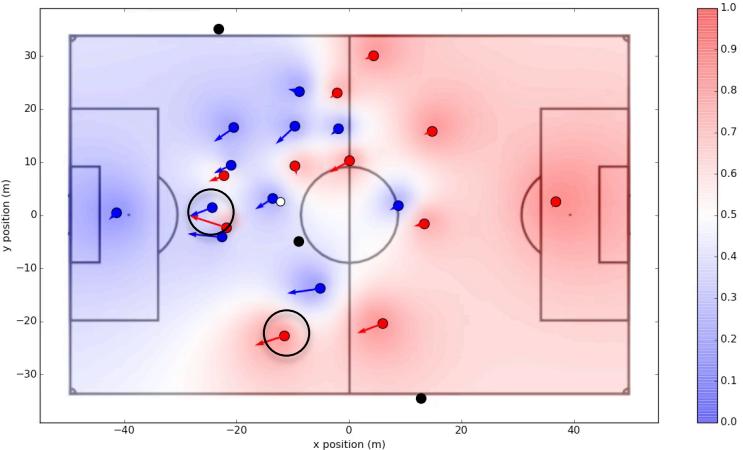
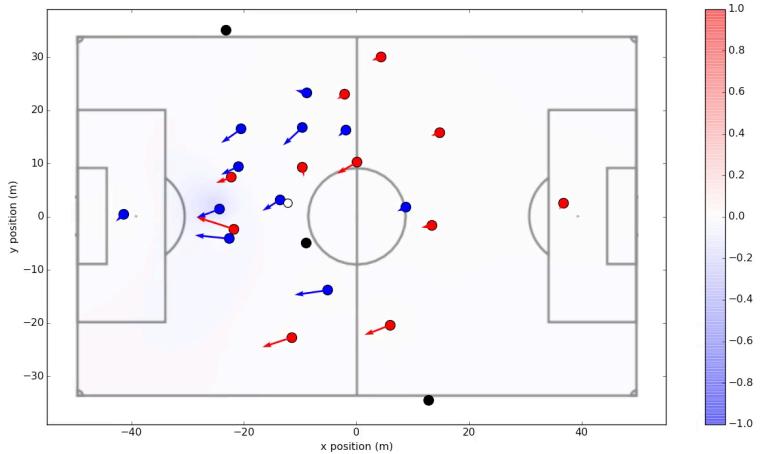
- Above we see the fitted beta value for specific midfielders when compared to the league average.
- Some are more controlling while others exert less impact.
- Error bars show the standard deviation among four games.

# Applications: Player Positioning

## Quantify the Effect of Player Positioning

- How much impact does a player's position have on his team's control of the pitch?
- In other words, how different would the PCF be if a certain player weren't on the pitch?

$$\Delta PCF(j, t_i, l_i) = \left[ \frac{\sum_i l_i t_i^{-\beta_i}}{\sum_i t_i^{-\beta_i}} - \frac{\sum_{i \neq j} l_i t_i^{-\beta_i}}{\sum_{i \neq j} t_i^{-\beta_i}} \right] / 2$$



# Thank you

William Spearman

#OptaProForum

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