## Using Models to Evaluate Statistics

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Using Models to Evaluate Statistics

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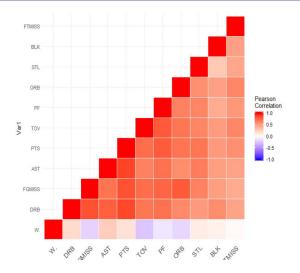
- Measuring Efficiency requires the aggregation of different indicators
- The crucial issue in aggregation is weighting

• possible solution : aggregate statistics with unit weights.

$$EFF = PTS + REB + AST + STL + BLK$$
  
 $-FGMISS - FTMISS - TOV$ 

- The formula combines statistics but the weights are not convincing
  - Why does a Missed FT have the same value with a Missed FG ?
  - are an assist and a missed free throw of equal value?

## NBA Stats and Winning Percentage



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- Regressing WINS on all indicators and construct a measure of the impact on each statistics using the regression coefficients
- the empirical model undergoes a serious danger of **over-parameterization** 
  - there are many indicators
  - they comove
  - and there is no strong correlation between NBA statistics and Winning percentage
- Regressing WINS on the NBA efficiency indicator
  - the empirical model undergoes a serious danger of **under-parameterization**

- We use the number of WINS in a regular season as the measurable counterpart of performance
- We theorize that the key concept to determine peformance is how efficiently teams use **possession**
- A possession starts when one team gains control of the ball and ends when that team gives it up (in other words, an offensive rebound would start a new play, not a new possession). Possession totals are guaranteed to be approximately the same for the two teams in a game.
- WINS depend on how efficiently a team **uses a possession** and on how costly it is to **acquire a possession**

• Construct an empirical counterpart for employed possession and acquired possession

$$EP_{i,t} = FGA_{i,t} + 0.45 * FTA_{i,t} + TOV_{i,t} - ORB_{i,t}$$

 $AP_{i,t} = OTOV_{i,t} + DRB_{i,t} + TEAMR_{i,t} + OFG_{i,t} + 0.45 * OFT_{i,t}$ 

Team Rebounds are not available from the NBA website, but we can construct them under the null that  $EP_{i,t} = AP_{i,t}$ 

$$FGA_{i,t} = OFG_{i,t} + 0.45 * OFT_{i,t} + OTOV_{i,t} + ORB_{i,t} + DRB_{i,t} + TEAMR_{i,t} - TOV_{i,t} - 0.45 * FTA_{i,t}$$

we know that a plausible estimate for *x* and *z* is 0.45, so we can get  $TRB_{i,t}$  as follows:

$$TEAMR_{i,t} = FGA_{i,t} - OFG_{i,t} - 0.45 * OFT_{i,t} - OTOV_{i,t} - ORB_{i,t} - DRB_{i,t} + TOV_{i,t} + 0.45 * FTA_{i,t}$$

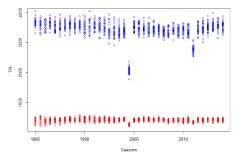
Define Field Goal Attempt Differenced as follows:

$$FGAD_{i,t} = FGA_{i,t} + TOV_{i,t} - ORB_{i,t} - OTOV_{i,t} - DRB_{i,t} - OFG_{i,t}$$

run the following regression

 $FGAD_{i,t} = \alpha + \beta_1 OFT_{i,t} + \beta_2 FTA_{i,t} + u_{it}$ we should have  $\beta_1 = 0.45, \beta_2 = -0.45, TRB_{i,t} = \alpha + u_{it}$ 

# Graphical Analysis

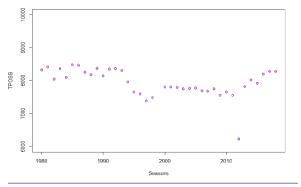


Estimated Team Rebounds

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## Graphical Analysis

Atlanta Hawks



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$$W_{it} = \beta_0 + \beta_1 \left( PTSxEP_{it} - PTSAxAP_{i,t} \right) + u_{it}$$
  
$$u_{it} \sim N.I.D \left( 0, \sigma^2 \right)$$

- Now we have a model, with estimates of all unknowns parameter and some related measure of uncertainty.
- Suppose you want to assess the impact on Wins of a specific statistic (say a three-point made shot)
- You can proceed via the following steps
  - Generate via the model a predicted value for wins in the case all statistics are kept at their average. This is called the baseline scenario simulation.
  - Generate via the model a predicted value for wins in case all the statistics are kept at their average except the one in whose effect you want to evaluate.
  - the difference gives the impact of the stats on WINs and its distribution
- Note that the model takes all feedbacks into account: one more 3points made gives you 3points at the cost of employing a possession

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- in the deterministic model simulation we do not aknowledge the importance of uncertainty in all the estimated equations in the model.
- Stochastic simulation fixes this
- coefficients are drawn from their distribution (Monte-Carlo simulation)
- Then artificial values are generated under two scenarios
  - a baseline scenario in which no exogenous variable (stats) is changed
  - an alternative scenario in which one of the stats is changed
  - the difference gives the impact of the stats on WINs and its distribution
- This exercise can be replicated N times (and therefore a Monte-Carlo simulation generates a vector of computer N simulated wins under the two scenarios and their difference).

Given the estimation of the model

$$W_{it} = \beta_0 + \beta_1 \left( PTSxEP_{it} - PTSAxAP_{i,t} \right) + u_{it}$$

We can simulate stochastically the effect of modification in efficiency on total expected wins by simulating:

$$W_{it} = \hat{\beta}_0 + \left(\hat{\beta}_1 + \sigma_{\hat{\beta}_1} v_{it}^t\right) \left(PTSxEP_{it} - PTSAxAP_{i,t}\right)$$

or we can simulate stochastically the distribution of wins

$$W_{it} = \beta_0 + \beta_1 \left( PTSxEP_{it} - PTSAxAP_{i,t} \right) + u_{it}$$

• the two simulations are very different

- model can be completed by adding auxiliary equations to model specific variables
  - Personal Foul and Blocked Shots

### The Values of NBA Statistics in Terms of Wins

Various Statistics Tracked for Players and Teams	If each variable increased by one, and nothing else changed, wins would change by	If each variable increased by 100, and nothing else changed, wins would change by
SCOR	ING STATISTICS	
Three-point field goals made	+0.066	+6.6
Opponent's three-point field goals made	-0.066	-6.6
Two-point field goals made	+0.033	+3.3
Opponent's two-point field goals made	-0.032	-3.2
Free throws made	+0.018	+1.8
Opponent's free throws made	-0.018	-1.8
Missed field goals	-0.034	-3.4
Missed free throws	-0.015	-1.5
POSSE	SSION STATISTICS	
Offensive rebounds	+0.034	+3.4
Turnovers	-0.034	-3.4
Defensive rebounds	+0.034	+3.4
Team rebounds	+0.034	+3.4
Opponent's turnovers	+0.034	+3.4
Steals	+0.034	+3.4
PERSONAL FO	ULS AND BLOCKED SHO	TS
Personal fouls	-0.018	-1.8
Blocked shots	+0.021	+2.1

TABLE 6.5 The Value of Various NBA Statistics in Terms of Wins

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