# LS88: Sports Analytics

Dean Oliver's Four Factor Basketball Model

An empirically driven model that explains basketball performance

It aims to answer "How do teams win basketball games?"

It features four fundamental components of the sport

- → Score efficiently
- → Protect the basketball on offense
- → Grab as many rebounds as possible
- → Get to the foul line and make FTs as often as possible

The Model

Team Performance = 
$$.4 \times Z(eFG\% - eFG\%_{\mathrm{Opp}})$$
  
 $- .25 \times Z(\mathrm{Turnover\%} - \mathrm{Turnover\%}_{\mathrm{Opp}})$   
 $+ .2 \times Z(OREB\% - OREB\%_{\mathrm{Opp}})$   
 $+ .15 \times Z(\mathrm{FT\ Rate} - \mathrm{FT\ Rate}_{\mathrm{Opp}})$ 

What are the Zs? What are the weights?

#### The outline

- → What is a factor model?
- → How are they used? They're ubiquitous in finance
- → How do we compute the Four Factor Model?
- → What does it tell us about basketball?

### **Factor Models**

First, what is a "factor model"?

#### Basic idea:

- → You have have observed outcomes (stock returns, team performance, etc)
- → There are "factors" or features or just generally quantities that drive performance of your outcomes
- → Typically you take the factor model as linear

Outcome = 
$$\beta_1 \cdot \text{Factor } 1 + \cdots + \beta_K \cdot \text{Factor } K$$

→ Common in Pyschology and Finance (I'll talk about finance)

### **Factor Models**

Quick aside:

LWTS was a linear model but not a factor model We were just estimating the run values of events

#### Factors instead would be things like:

- → Measures of getting on-base or power
- → Measures of baserunning aggressiveness
- → Measures of pitching or defense

We use a factor model to explain market phenomena For example asset prices/returns, or correlation of returns

The most important factor: the "market"

- → The notion of the market is a bit abstract
- → But we can substitute something for the market though. For example DJIA or S&P500 indices
- → A model for stock returns (the famed CAPM)

Stock Return = Expected Return +  $\beta$  · Market Return + Idiosyncratic Return

Model for stock returns (the famed CAPM)

Stock Return = Expected Return +  $\beta$  · Market Return + Idiosyncratic Return

The "market" explain/drives the variation in returns you observe (and after that, it's all just noise)

Stocks go up/down with the market so every stock is correlated

Model for stock returns (the famed CAPM)

Stock Return = Expected Return +  $\beta$  · Market Return + Idiosyncratic Return

Magnitude of idiosyncratic return and beta: fraction of variation due to the market

Beta > 1: stock is extra reactive to market

Beta < 1: stock not as reactive

Beta = 0: stock not affected by market

Beta < 0: stock anti-correlated with market

Another (potential) factor: momentum

Momentum is recent historical performance. Trending up? Trending down?

What do you do with the factors? Try to build portfolios based on them:

- → Try to match the market return but minimize risk
- → Place a bet on momentum
- → Anything you want: you look at the "betas" and decide how you want the portfolio to exposed to the risk of the factors

So what can be a factor?

Potentially anything you want. But you might concoct a bogus factor which doesn't explain anything about the market (and potentially lose a lot of money!!)

What else?

- → Macroeconomic factors: economy level data like employment or interest rates
- → "Fundamental" factors: things like value (price/earnings ratio) or momentum
  - Anything related to the company
- → Statistical: not firm or directly observable
  - Observed in data but nothing fundamental/macro explains it

What about the weights? How do we determine the importance of factors?

Ultimately, we have to figure out on our own how to weight the factors for each stock

→ Market factor is the most important for stocks (not bonds). But how important?

The short answer: regression

So what are Dean Oliver's Four Factors? We know the intuitive ideas from basketball:

- → Efficient shooting
- → Turnovers
- → Rebounding
- → Free Throws

These will be fundamental factors: compute values from team observables

Okay, we know the concepts, but what should we actually compute?

#### Shooting:

$$eFG\% = \frac{FG + .5 \cdot 3FG}{FGA}$$

#### Turnovers:

Turnover Rate = 
$$\frac{\text{Turnovers}}{\text{Possessions}}$$

#### Rebounding:

$$OREB\% = \frac{Off Reb}{Off Reb + Opposition Def Reb}$$

#### Free Throws:

$$FT Rate = \frac{FT}{FGA}$$

or FT Rate = 
$$\frac{FTA}{FGA}$$

Okay, so now what?

- → We want to explain team performance
- → Dean Oliver suggests weighting the factors by importance: 40, 25, 20, 15

Okay... so we just multiply each of those values by those numbers and add them up?

None of those numbers are related at all: different avg level, different variation

## **Z-Scores and Standardizing Data**

A bit of an important digression for a very useful and great tool for standardizing data

Suppose you have some variables on different levels and different variations We do. In fact, this happens basically all the time.

The goal is to standardize the variables

Remove average level, inherent variation, and units

## **Z-Scores and Standardizing Data**

You can quantify effects in terms of standard units like *standard deviation* 

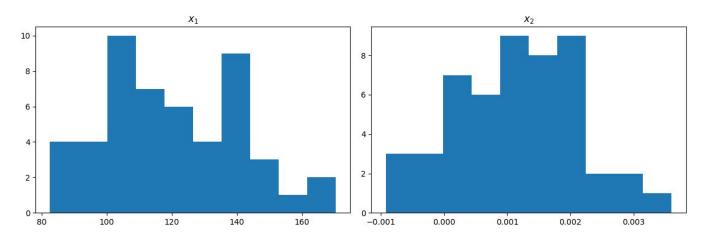
- → An example: two variables see a change of .01 and 100, respectively. Two totally different levels of change
- → But if the standard deviation for those variables are .01 and 100, then it's really the same level of change

For data observations  $X_1, ... X_N$ , the Z-Scores are given by

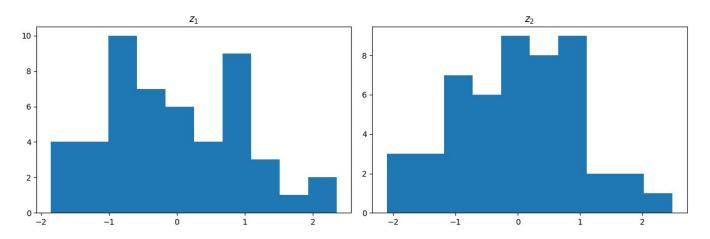
$$Z_i = \frac{X_i - \text{Avg of } X_1, \dots, X_N}{\text{Std Dev of } X_1, \dots, X_N}$$

The mechanism is to de-mean the data and then scale it so it has Std Dev = 1

#### Non-standardized data:



#### Standardized data:



Histograms have the same shape: data hasn't fundamentally change Just shifted and scaled

Okay, so now everything is on the same scale

→ 1.1 means the same thing for each factor

Coefficients have meaning: they now indicate a relative importance

And now we can just use Dean Oliver's suggested weightings

But wait, one more issue: there are 8 values for the team and its opposition

We take the raw components, compute the difference, and then compute Z-scores

```
Shooting Factor = Z(eFG\% - eFG\%_{\mathrm{Opp}})

Turnover Factor = Z(\mathrm{Turnover}\% - \mathrm{Turnover}\%_{\mathrm{Opp}})

Rebounding Factor = Z(OREB\% - OREB\%_{\mathrm{Opp}})

Free Throw Factor = Z(\mathrm{FT} \; \mathrm{Rate} - \mathrm{FT} \; \mathrm{Rate}_{\mathrm{Opp}})
```

The final model

Team Performance = 
$$.4 \times Z(eFG\% - eFG\%_{\mathrm{Opp}})$$
  
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Turnovers are bad hence the negative sign

So where did those weightings come from?

Honestly, I don't know and I haven't seen anyone offer an explanation

However, I have seen evidence that shooting should be weighted more heavily

We will revisit this when doing statistical inference

Anyhow, let's look at the data and how this all works

Player performance is about efficient production using possessions

Our goal should be to rate players comprehensively

We know how to evaluate their scoring efficiency

What about the rest of the play?

Enter PER: Player Efficiency Rating

- → Developed by John Hollinger in the late 90s
- → Based entirely on well known box score stats: FGs, FTs, Steals, Rebs, etc.
- → What PER tries to do:

A comprehensive, model-based rating that counts positive and negative contributions and rates them on a per possession basis.

$$uPER = \frac{1}{min} \times \left($$

Three Pointers Made

- + Contributions from Assists
- + Contributions from FGs
- + Contributions from FTs
- Contributions from TOs
- Contributions from Missed FGs
- Contributions from Missed FTs
- + Contributions from Def Rebounds
- + Contributions from Off Rebounds
- + Contributions from Steals
- + Contributions from Blocks
- Contributions from Fouls

- → Raw *uPER* is a per-minute efficiency calculation
- → We compute the final PER by first adjusting for the pace of play of teams
- → And finally adjusting to league average

  Performance is put relative to the league average

We first need to understand and deconstruct PER

Then we can get into the pros and cons and finding potentially better approaches

Let's study PER hands-on in a lab...