

Rating, Ranking, Betting and Mathematics

R VITTAL RAO

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Mark: “No, I need your algorithm to rank chess players”

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Mark: "Yeah"

Ed: "Do you think it is such a good idea?"

Mark: "I need your algorithm - I need your algorithm"

Ed: "With each girl's base rating 1400..."

Ed proceeds to write the formulae (on a window with a crayon)

Eduardo's Formula!

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And all those expectations are expressed this way”

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Jesse Eisenberg as Mark Zuckerberg (founder of Facebook)

Andrew Garfield as Eduardo Saverin (cofounder)

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Jesse Eisenberg as Mark Zuckerberg (founder of Facebook)

Andrew Garfield as Eduardo Saverin (cofounder)

Justin Timberlake as Sean Parker (cofounder)

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RATING and RANKING

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Every Rating creates a Ranking by arranging the ratings in descending order

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Such pairwise comparisons are at the heart of most of the methods of rating and ranking

General Mathematical Techniques

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Most of the rating system use one or more of the following:

- ▶ Linear Algebra - Linear Systems of Equations, eigenvalues and Eigenvectors)

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- ▶ Game Theory

ARROW'S IMPOSSIBILITY THEOREM

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(*Non – dictatorship*)

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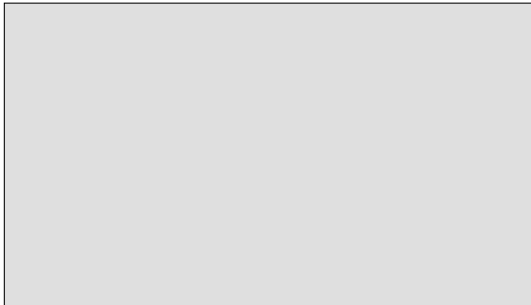
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Modified version used in **BCS (BOWL CHAMPIONSHIP SERIES)**

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$$\begin{aligned} r_i &= \frac{1 + k}{2 + 2k} \\ &= \frac{1}{2} \end{aligned}$$

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2012:

On June 26, 2012, it was announced that the Bowl Championship Series will be replaced by a four-team playoff, effective for the 2014-15 season to be known as the College Football Playoff.

Elo's Rating

Chess Rating

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How are $(r_i)_{old}$ and $(r_i)_{new}$ related?

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For chess games,

$$S = \begin{cases} 1 & \text{if } P_i \text{ wins} \\ \frac{1}{2} & \text{if } P_i \text{ draws} \\ 0 & \text{if } P_i \text{ loses} \end{cases}$$

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$$S_{ij} = \begin{cases} 1 & \text{if } P_i \text{ beats } P_j \\ \frac{1}{2} & \text{if game is a draw} \\ 0 & \text{if } P_i \text{ loses} \end{cases}$$

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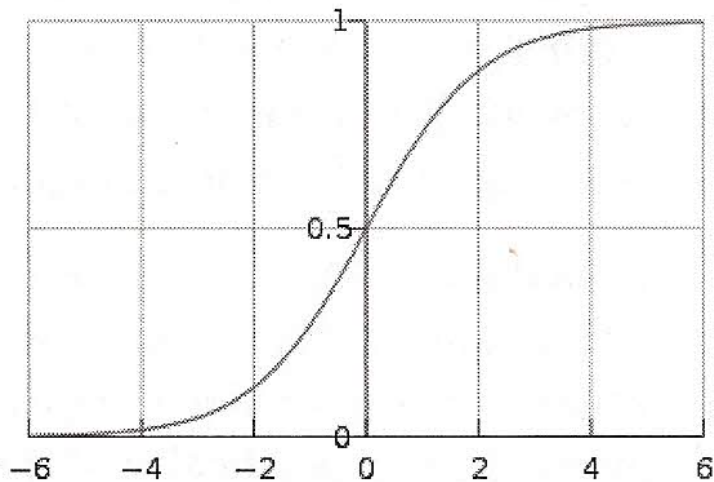
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The Logistic Curve $L(x)$



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$$\begin{aligned} \mu_{ij} + \mu_{ji} &= \frac{1}{1 + 10^{-\alpha_{ij}}} + \frac{1}{1 + 10^{\alpha_{ij}}} \\ &= 1 \end{aligned}$$

The Total score is One

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Clearly we have

$$S_{ij} + S_{ji} = 1$$

The Total Rating is Invariant

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If T_i and T_j play then only the ratings of T_i and T_j change after the game depending on the reward they get. Hence

$$\sum_{k=1}^n (r_k)_{new} =$$

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SPREAD BETTING

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He started the new method of trading and changed the way people bet

Beat The Spread

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Since the time of McNeil,

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"BEATING THE SPREAD"

Beat The Spread

Since the time of McNeil,
“BEATING THE SPREAD
is the HOLY GRAIL of the betting world

What does “Beating the spread mean?”

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Suppose \mathcal{I} bet on A

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In Case 1, \mathcal{I} win the bet because A beats B by more than 10 Points

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In Case 1, \mathcal{I} win the bet because A beats B by more than 10 Points

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My Bet on B

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Suppose I bet on B

My Bet on B

Suppose I bet on B

In Case 1 \mathcal{I} lose because A has beaten B by more than 10 points

My Bet on B

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In Case 3 \mathcal{I} win because anyway B has won

Handling Ties

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To handle ties the book makers usually give the spread as 7.5 or 8.5 etc.

Using the Spread For Rating

The Spread matrix

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The Spread Matrix S :

$S = (S_{ij})_{n \times n}$ (diagonal entries are defined to be zero)

S is a SKEW SYMMETRIC MATRIX

The Rating Spread matrix

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$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

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The Rating Spread Matrix:

$\mathcal{X} = (x_{ij})_{n \times n}$ where $x_{ij} = x_i - x_j$

Ideal Situation

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The best way is to find the error $\mathcal{S} - \mathcal{X}$ and find “the” \mathbf{x} that minimizes this error

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So get a rating \mathbf{r} such that

$$\|\mathcal{S} - \mathcal{R}\| < \|\mathcal{S} - \mathcal{X}\| \text{ for any other rating } \mathbf{x} \neq \mathbf{r}$$

Can we find such an r ?

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A little bit of simple Linear Algebra and calculus shows that such an r exists and is given by

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Conclusion

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Only he knows!!

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(Nor rated by heartless mathematicians' formulae!!)

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Beauty and the Beast

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Arrow: “Beastly politicians cannot be ranked”

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