# An Intransitivity Model for Matchup and Pairwise Comparison 

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#### Abstract

Objectives - Predict pairwise matchup. - Unify several existing models on representation learning.

\section*{Intransitivity}

An intuitive example a rock-paper-scissors game, the pairwise matchup result is judged by three rules: \{paper $\succ$ rock, rock $\succ$ scissors and scissors $\succ$ paper \}. A transitive model results in a transitive dominance paper $\succ$ scissors, that violates the third rule scissors $\succ$ paper. Examples sports tournaments; online games; election process.


## Related works

## Problem setting

- Candidate players $i$ and $i \in \mathbf{P}$ with $|\mathbf{P}|=N$
-4-tuples $\left(i, j, n_{i}, n_{j}\right)$, and $i \succ j:=(i, j, 1,0)$.
- Matchup matrix $\mathbf{M} \in \mathcal{R}^{N \times N}$.
- $M_{i j}>0$ means item $i$ has a comparative advantage over item $j$

1. Bradley-Terry Model (Bradley et al., 1952)

$$
\begin{equation*}
\operatorname{Pr}(i \succ j)=\frac{\exp \left(\gamma_{i}\right)}{\exp \left(\gamma_{i}\right)+\exp \left(\gamma_{j}\right)}=\frac{1}{\left.1+\exp \left(-M_{i j}\right)\right)}, \tag{1}
\end{equation*}
$$

where $\gamma_{i}$ is the ability of winning for player $i$.

$$
M_{i j}=\gamma_{i}-\gamma_{j} .
$$

2. Blade-Chest-Inner Model (Chen et al., 2016a)

$$
\begin{equation*}
M_{i j}=\mathbf{x}_{i}^{\text {blade }}{ }^{\top} \mathbf{x}_{j}^{\text {chest }}-\mathbf{x}_{j}^{\text {blade }} \mathbf{x}_{i}^{\text {chest }}+\gamma_{i}-\gamma_{j}, \tag{2}
\end{equation*}
$$

3. Neural network framework of Blade-Chest (Chen et al., 2016b)

Top layer is the blade-chest-inner model (2):

$$
\operatorname{Pr}(i \succ j \mid g)=\sigma(M(i, j \mid g)) .
$$

4. Blade-Chest-Sigma Model (Duan et al., 2017)

$$
\begin{equation*}
M_{i j}=\mathbf{x}_{i}^{\top} \Sigma \mathbf{x}_{j}+\mathbf{x}_{i}^{\top} \Gamma \mathbf{x}_{i}-\mathbf{x}_{j}^{\top} \Gamma \mathbf{x}_{j}, \tag{3}
\end{equation*}
$$

where $\Sigma, \Gamma \in \mathcal{R}^{d \times d}$ are the transitive matrices.

## Proposed method

- A generalized model with more expressiveness by a low-rank matrix.
- Neural Network framework of the generalized model.
- A quantitative evaluation of the existence of intransitivity in datasets.


## Matchup function

Define the representation matrix as

$$
\begin{gathered}
\mathbf{X}^{\text {blade }}=\left(\mathbf{x}_{1}^{\text {blade }}, \ldots, \mathbf{x}_{N}^{\text {blade }}\right), \mathbf{X}^{\text {chest }}=\left(\mathbf{x}_{1}^{\text {chest }}, \ldots, \mathbf{x}_{N}^{\text {chest }}\right), \boldsymbol{\gamma}=\left(\gamma_{1}, \gamma_{2}, \ldots, \gamma_{N}\right) . \\
(2) \Rightarrow \mathbf{M}=\mathbf{X}^{\text {blade }}{ }^{\top} \mathbf{X}^{\text {chest }}-\mathbf{X}^{\text {chest }} \mathbf{X}^{\text {blade }}+\boldsymbol{\gamma}^{\top} \mathbf{1}-\mathbf{1}^{\top} \boldsymbol{\gamma} .
\end{gathered}
$$

Replace the matrix product $\mathbf{X}^{\text {blade }}{ }^{\top} \mathbf{X}^{\text {chest }}$ by a new matrix $\mathbf{Y}$ as

$$
\mathbf{Y}=\mathbf{X}^{\text {blade }}{ }^{\top} \mathbf{X}^{\text {chest }},
$$

which results in a general representation of the matchup matrix as

$$
\begin{equation*}
\mathbf{M}=\left(\boldsymbol{\gamma}^{\top} \mathbf{1}-\mathbf{1}^{\top} \boldsymbol{\gamma}\right)+\left(\mathbf{Y}-\mathbf{Y}^{\top}\right) \text { s.t. } \operatorname{rank}(\mathbf{Y}) \leq D . \tag{4}
\end{equation*}
$$

Objection (assuming $i$ is the winner; same with MSE loss)

$$
\begin{equation*}
\Theta \Sigma_{(i, j)} \log \operatorname{Pr}(i \succ j \mid \Theta) . \tag{5}
\end{equation*}
$$

Top layer the generalized intransitivity model (4):

$$
\operatorname{Pr}(i \succ j)=\sigma\left(M_{i j}\right)=\sigma\left(Y_{i j}-Y_{j i}\right) .
$$



Figure: An illustration of the proposed generalized intransitivity framework

## Properties

1. $\mathbf{Y}-\mathbf{Y}^{\top}$ can represent an arbitrarily complex matchup matrix by removing the rank constraint.
2. It is equivalent to the BT model when $\operatorname{rank}(\mathbf{Y})=0$.
3. It can represent the intransitivity when $\operatorname{rank}(\mathbf{Y})=1$ :

$$
\mathbf{Y}=\left(x_{1}^{\text {blade }}, x_{2}^{\text {blade }}, \ldots, x_{N}^{\text {blade }}\right)^{\top}\left(x_{1}^{\text {chest }}, x_{2}^{\text {chest }}, \ldots x_{N}^{\text {chest }}\right),
$$

the matchup matrix without the strength terms becomes

$$
M_{i j}=x_{i}^{\text {blade }} x_{j}^{\text {chest }}-x_{i}^{\text {chest }} x_{j}^{\text {blade }} .
$$

Assume that $i \succ j$ and $j \succ k$ (i.e., $M_{i j}>0$ and $M_{j k}>0$ ), then taking $x_{i}^{\text {chest }}>0, x_{j}^{\text {chest }}<0$, and $x_{k}^{\text {chest }}>0$ shows $k \succ i\left(\right.$ i.e., $\left.M_{i k}<0\right)$.

## Datasets and Experiments

- Intrans.: whether the dataset contains intransitivity
- No.IntPlayer: the fraction of players invovled in rock-paper-scissors
- Int.Ratio: the amount of the rock-paper-scissors-like relationship (\%)

Table: Summary of the real-world datasets

| Dataset | Players |  |  | Records | Intrans. |
| :---: | :---: | :---: | :---: | :---: | :---: | No.IntPlayer Int.Ratio 0

How does the proposed method perform?
Table: Test accuracy on the real-world datasets

| Dataset | Bradley-Terry | Blade-Chest-Inner | Blade-Chest-Sigma | Neural BC | Proposed model |
| :---: | :---: | :---: | :---: | :---: | :---: |
| SushiA | $0.6525 \pm 0.0011$ | $0.6546 \pm 0.0006$ | $0.6560 \pm 0.0004$ | $0.6630 \pm 0.0004$ | $\mathbf{0 . 6 6 3 2} \pm \mathbf{0 . 0 0 0 3}$ |
| SushiB | $0.6257 \pm 0.0025$ | $0.6235 \pm 0.0150$ | $0.6414 \pm 0.0019$ | $0.6561 \pm 0.0017$ | $\mathbf{0 . 6 5 6 3} \pm \mathbf{0 . 0 0 1 1}$ |
| MovieLens100K | $0.6785 \pm 0.0005$ | $0.6792 \pm 0.0004$ | $0.6789 \pm 0.0003$ | $0.6950 \pm 0.0019$ | $\mathbf{0 . 6 9 7 3} \pm \mathbf{0 . 0 0 0 2}$ |
| Election A5 | $0.6478 \pm 0.0017$ | $0.6489 \pm 0.0011$ | $0.6494 \pm 0.0018$ | $0.6550 \pm 0.0030$ | $\mathbf{0 . 6 5 6 0} \pm \mathbf{0 . 0 0 1 8}$ |
| Election A9 | $0.6028 \pm 0.0003$ | $0.6096 \pm 0.0007$ | $0.6047 \pm 0.0008$ | $0.6174 \pm 0.0003$ | $\mathbf{0 . 6 1 7 5} \pm \mathbf{0 . 0 0 0 3}$ |
| Election A17 | $0.5189 \pm 0.0001$ | $0.5305 \pm 0.0010$ | $0.5296 \pm 0.0013$ | $0.5582 \pm 0.0003$ | $\mathbf{0 . 5 5 9 8} \pm \mathbf{0 . 0 0 0 2}$ |
| Election A48 | $0.5993 \pm 0.0001$ | $0.6001 \pm 0.0001$ | $0.5996 \pm 0.0001$ | $0.6060 \pm 0.0001$ | $\mathbf{0 . 6 0 5 6} \pm \mathbf{0 . 0 0 0 1}$ |
| Election A81 | $0.6013 \pm 0.0001$ | $0.6018 \pm 0.0001$ | $0.6011 \pm 0.0002$ | $0.6194 \pm 0.0001$ | $\mathbf{0 . 6 1 9 4} \pm \mathbf{0 . 0 0 0 1}$ |
| SF4-5000 | $0.5079 \pm 0.0078$ | $0.5181 \pm 0.0171$ | $0.5358 \pm 0.0049$ | $0.5514 \pm 0.0008$ | $\mathbf{0 . 5 4 9 6} \pm \mathbf{0 . 0 0 2 1}$ |
| DotA | $0.6334 \pm 0.0077$ | $0.6432 \pm 0.0034$ | $0.6420 \pm 0.0051$ | $0.6468 \pm 0.0031$ | $\mathbf{0 . 6 4 8 5} \pm \mathbf{0 . 0 0 2 5}$ |
| Pokemon | $0.8157 \pm 0.0094$ | $0.8495 \pm 0.0016$ | $0.8187 \pm 0.0168$ | $0.8943 \pm 0.0040$ | $\mathbf{0 . 8 9 4 9} \pm \mathbf{0 . 0 0 2 1}$ |

## References

Bradley et al., 1952, Ran analysis of incomplete block designs: I. the method of paired comparisons. Chen et al., 2016a, Modeling intransitivity in matchup and comparison data.
Chen et al., 2016b, Predicting matchups and preferences in context.
Duan et al., 2017, A Generalized Model for Multidimensional Intransitivity.

