

Lifted Relational Team Embeddings for Predictive Sport Analytics

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The Problem

- predicting future match outcome from historical data
- soccer matches results from EPL 2004–2016
- no additional information

Date	Home	Away	Score H	Score A
10/6/2004	Arsenal	Chelsea	3	1
...
11/12/2016	Bolton	Everton	2	2

Knowledge Representation

Predicate	Description
$\text{home}(Tid)$	Team Tid is home team w.r.t. prediction match.
$\text{away}(Tid)$	Team Tid is away team w.r.t. prediction match.
$\text{team}(Tid, name)$	Team Tid has name $name$.
$\text{win}(Mid, Tid_1, Tid_2)$	Win of home team Tid_1 over team Tid_2 in match Mid .
$\text{draw}(Mid, Tid_1, Tid_2)$	Draw between home team Tid_1 and Tid_2 in match Mid .
$\text{loss}(Mid, Tid_1, Tid_2)$	Loss of home team Tid_1 to team Tid_2 in match Mid .
$\text{scored}(Mid, Tid, n)$	The team Tid scored more than n goals in match Mid .
$\text{conceded}(Mid, Tid, n)$	The team Tid conceded more than n goals in match Mid .
$\text{goal_diff}(Mid, n)$	Difference in goals scored by the teams is greater than n .
$\text{recency}(Mid, n)$	The match Mid was played more than n rounds ago.

Lifted Relational Neural Networks

- framework utilizing a fragment of relational fuzzy logic
- parameter training by gradient descend
- model = lifted template for neural network
- LRNN grounds the template w.r.t. the different examples
- different computational graph for each example

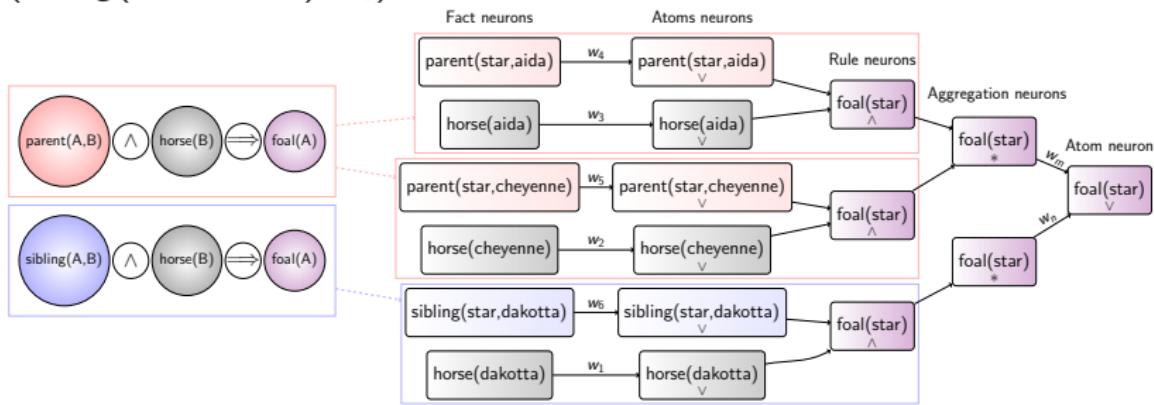
LRNN Toy Example

Rules:

$$(w_m : foal(A) \leftarrow parent(A, B) \wedge horse(B)),$$
$$(w_n : foal(A) \leftarrow sibling(A, B) \wedge horse(B)),$$

Facts:

$$(horse(dakotta), w_1), (horse(cheyenne), w_2),$$
$$(horse(aida), w_3), (parent(star, aida), w_4), (parent(star, cheyenne), w_5),$$
$$(sibling(star, dakotta), w_6)$$



Embedding Layer

Embedding declaration:

$$w_1^{(0)} : type_1(T) \leftarrow team(T, arsenal)$$
$$w_2^{(0)} : type_2(T) \leftarrow team(T, arsenal)$$
$$w_3^{(0)} : type_3(T) \leftarrow team(T, arsenal)$$

...

$$w_j^{(0)} : type_3(T) \leftarrow team(T, everton)$$

Predictive rules:

$$w_{(1;1)}^{(1)} : outcome \leftarrow home(T1) \wedge type_1(T1) \wedge away(T2) \wedge type_1(T2).$$
$$w_{(1;2)}^{(1)} : outcome \leftarrow home(T1) \wedge type_1(T1) \wedge away(T2) \wedge type_2(T2).$$

...

$$w_{(3;3)}^{(1)} : outcome \leftarrow home(T1) \wedge type_3(T1) \wedge away(T2) \wedge type_3(T2).$$

Relational Extension

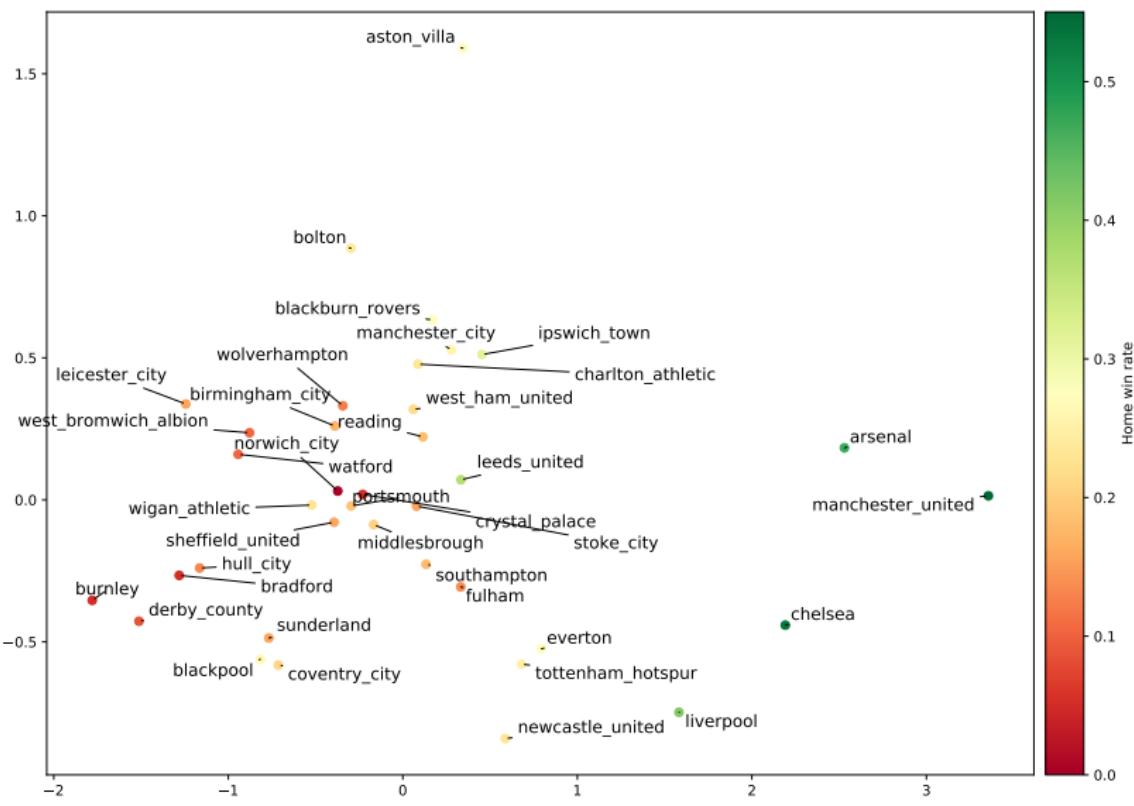
Extension:

- $w_1^{(2)} : outcome(M, H, A) \leftarrow win(M, H, A)$
- $w_2^{(2)} : outcome(M, H, A) \leftarrow draw(M, H, A)$
- $w_3^{(2)} : outcome(M, H, A) \leftarrow loss(M, H, A)$

Predictive rules:

- $w_{h-h(1;1)}^{(1)} : outcome \leftarrow home(T1) \wedge type_1(T1) \wedge outcome(M, T1, T2) \wedge type_1(T2).$
- $w_{h-a(1;1)}^{(1)} : outcome \leftarrow home(T1) \wedge type_1(T1) \wedge outcome(M, T2, T1) \wedge type_1(T2).$
- $w_{h-h(1;2)}^{(1)} : outcome \leftarrow home(T1) \wedge type_1(T1) \wedge outcome(M, T1, T2) \wedge type_2(T2).$
- ...
- $w_{a-a(3;3)}^{(1)} : outcome \leftarrow away(T1) \wedge type_3(T1) \wedge outcome(M, T2, T1) \wedge type_3(T2).$

Embeddings



Comparison with SotA

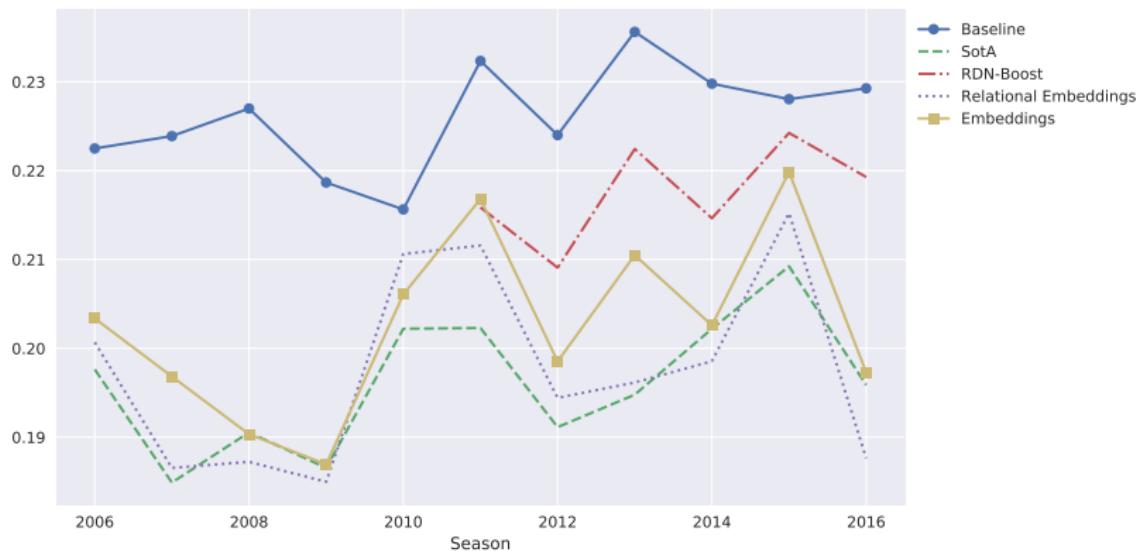


Figure: Comparison of performance of the learners on English Premier League as measured by the RPS metric (lower is better).

Conclusion

- promising preliminary results
- easily applicable to different sports
- extensible with more information (goals scored, match recency, ...)
- natural incorporation of domain knowledge

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Thank you for your attention.