

Neural Regression with Embeddings for Numeric Attribute Prediction in Knowledge Graphs - Appendix

No Author Given

No Institute Given

1 Experiment Details

All embedding models for link prediction tasks are trained with embedding dimension 32, for 256 epochs, using a learning rate of 0.05 and a mini-batch size of 1024 using the KvsAll scoring technique for standard as well as combined training with LITEM. Similarly, the LITEM is trained with the same embedding dimension as KGE models. We use a dropout of 0.3, and a learning rate of 0.01 and train the LITEM for 200 epochs for literal-prediction tasks using pre-trained embeddings.

2 Literal Embedding Model

Algorithm 1 Literal Embedding Model

Input: Attribute triples $\{(e, a, v)\} \in \mathcal{G}$, Learning rate α , Number of epochs n ,
Pre-trained entity embeddings $\mathbf{E} \in \mathbb{R}^{|\mathcal{A}| \times d}$,

Initialize $\theta = \{ \text{attribute embeddings } \mathbf{A} \in \mathbb{R}^{|\mathcal{A}| \times d}, \text{ weight matrices } \mathbf{W}_1, \mathbf{W}_2, \text{ and} \}$
biases $\mathbf{b}_1, \mathbf{b}_2 \}$ \triangleright All learnable parameters are part of θ

Output: Trained parameters θ

for epoch = 1 **to** n **do**

$$\hat{y} = \mathbf{W}_2 \cdot (\mathbf{f}(\mathbf{W}_1 \cdot [\mathbf{e}, \mathbf{a}] + \mathbf{b}_1) + [\mathbf{e}, \mathbf{a}]) + \mathbf{b}_2$$

 Compute the MAE loss: $\mathcal{L} = \frac{1}{N} \sum (|\hat{y} - v|)$

 Compute the gradient $\nabla_{\theta} L$ of the loss with respect to θ

 Update the parameters: $\theta \leftarrow \theta - \alpha \nabla_{\theta} L$

end for

return θ

3 Literal Prediction

Table 1. Numeric Attribute Prediction performance of different embedding models with standard training on FB15k-237, YAGO15K and DB15K evaluated using Mean Absolute Error (MAE) Metric. All the embedding models are trained for 256 epochs, with a learning rate of 0.05 and 32-dimensional real-valued vectors with the KvsAll Scoring technique.

relation	CmplEx	DeCaL	DistMlt	Keci	MuRE	OMult	QMult	TransE
FB15k-237								
release_date	5.317	5.333	5.070	<u>5.022</u>	5.062	5.461	5.140	4.813
loc.date_founded	185.434	183.784	182.430	<u>174.937</u>	178.819	182.161	183.216	169.235
latitude	9.276	9.351	8.379	8.959	<u>7.573</u>	8.763	9.578	6.420
longitude	22.550	22.843	20.609	21.226	<u>20.439</u>	24.194	23.402	15.935
loc.area	1.67e6	1.68e6	1.67e6	1.68e6	<u>1.64e6</u>	1.70e6	1.67e6	1.64e6
org.date_founded	<u>58.269</u>	61.449	63.043	64.214	61.223	60.434	63.648	56.818
date_of_death	40.127	40.118	44.889	41.223	38.519	<u>37.994</u>	39.873	35.760
date_of_birth	22.860	23.716	24.065	22.837	<u>21.428</u>	22.923	22.906	20.675
height_meters	0.066	<u>0.067</u>	0.066	<u>0.067</u>	0.066	0.068	0.069	0.074
weight_kg	12.246	11.389	12.980	<u>11.290</u>	12.932	11.713	10.650	12.659
pop._number	5.09e6	5.44e6	4.79e6	5.03e6	<u>4.46e6</u>	5.57e6	4.79e6	4.39e6
YAGO15K								
diedOnDate	48.445	47.529	48.797	46.645	<u>46.045</u>	47.261	48.379	44.932
happenedOnDate	41.277	42.306	42.882	44.801	<u>36.921</u>	39.471	39.587	29.847
Latitude	7.416	7.216	6.972	6.939	<u>6.189</u>	7.245	6.815	4.652
Longitude	30.081	27.004	30.644	31.004	<u>26.406</u>	36.895	32.239	17.490
BornOnDate	20.690	19.798	20.525	20.776	<u>18.999</u>	20.061	20.516	17.458
CreatedOnDate	67.496	67.581	67.695	70.348	<u>62.844</u>	68.904	68.421	62.785
DestroyedOnDate	27.670	28.214	29.231	<u>27.093</u>	27.899	27.564	27.649	23.658
DB15K								
birthDate	18.318	18.598	18.306	17.619	<u>15.973</u>	18.943	17.691	14.403
completionDate	8.275	8.377	8.180	<u>7.259</u>	8.016	8.059	8.426	5.854
deathDate	20.115	21.330	21.550	20.036	20.876	21.349	18.595	<u>19.432</u>
formationDate	43.400	37.897	39.715	43.236	39.298	37.228	<u>38.110</u>	39.945
foundingDate	<u>39.780</u>	40.739	42.937	43.307	43.092	44.545	40.878	37.933
height	<u>1.798</u>	1.869	2.280	2.163	1.663	2.528	2.189	2.271
releaseDate	<u>11.319</u>	12.818	12.839	12.035	11.728	12.982	12.721	9.218
latitude	<u>8.077</u>	8.352	8.928	8.201	8.105	9.474	9.105	5.944
longitude	30.947	29.970	30.662	32.365	<u>29.804</u>	34.371	31.704	21.737

Table 2. Literal Prediction performance of different numeric attribute propagation approaches on FB15k-237 and YAGO15K compared with LitEm on Mean Absolute Error (values for KGA, NAP++ and MrAP) (Values taken from [1])

Relation	KGA	NAP++	MrAP	LitEm
FB15k-237				
date_of_birth	18.900	22.100	15.000	18.824
date_of_death	20.600	52.300	16.300	32.620
release_date	4.000	9.900	6.300	4.258
org.date_founded	49.000	59.300	58.300	46.741
loc.date_founded	76.000	92.100	98.800	150.873
latitude	2.100	11.800	1.500	5.221
longitude	7.100	54.700	4.000	11.325
area	6.10e4	4.40e5	4.40e5	1.65e6
population_number	4.00e6	7.50e6	2.10e7	4.87e6
height_meters	0.077	0.080	0.086	0.074
weight_kg	11.600	15.300	12.900	12.653
YAGO15K				
BornOnDate	16.300	23.200	19.700	15.586
diedOnDate	30.800	45.700	34.000	40.515
CreatedOnDate	58.200	83.500	70.400	57.081
DestroyedOnDate	23.300	38.200	34.600	21.436
happenedOnDate	29.900	73.700	54.100	26.064
Latitude	3.400	8.700	2.800	3.090
Longitude	7.200	43.100	5.700	11.506

4 Link Prediction

Link prediction performances of KGE models with and without combined training with the LitEm model on Mutag and FB15k-237 datasets.

Table 3. Link prediction performance of different embedding models with 32-dimensional real-valued vectors on the FB15k-237 dataset without and with combined training with LitEm. Each model’s results on training, test, and validation sets are shown in three rows; bold marks the best per-metric value for each model.

Models	KGE				KGE + LitEm			
	MRR	H@1	H@3	H@10	MRR	H@1	H@3	H@10
ComplEx	0.458	0.352	0.512	0.662	0.415	0.312	0.465	0.616
	0.247	0.174	0.270	0.393	0.234	0.165	0.253	0.372
	0.247	0.174	0.269	0.393	0.237	0.168	0.256	0.372
DeCaL	0.513	0.413	0.565	0.706	0.496	0.393	0.549	0.695
	0.262	0.186	0.286	0.415	0.260	0.183	0.283	0.413
	0.264	0.187	0.288	0.418	0.260	0.184	0.284	0.413
DistMult	0.475	0.372	0.529	0.676	0.454	0.347	0.509	0.660
	0.245	0.170	0.268	0.396	0.243	0.170	0.266	0.389
	0.247	0.173	0.270	0.394	0.243	0.170	0.265	0.388
Keci	0.543	0.438	0.602	0.744	0.519	0.415	0.576	0.718
	0.260	0.182	0.284	0.417	0.268	0.190	0.290	0.426
	0.264	0.186	0.289	0.420	0.270	0.191	0.296	0.427
OMult	0.530	0.431	0.584	0.721	0.468	0.366	0.518	0.665
	0.230	0.156	0.253	0.376	0.230	0.158	0.251	0.371
	0.233	0.160	0.255	0.378	0.231	0.159	0.252	0.374
QMult	0.544	0.443	0.598	0.736	0.535	0.434	0.589	0.728
	0.255	0.179	0.276	0.410	0.260	0.182	0.282	0.416
	0.261	0.184	0.283	0.413	0.262	0.185	0.283	0.417
TransE	0.429	0.306	0.499	0.660	0.409	0.287	0.476	0.638
	0.307	0.217	0.341	0.481	0.301	0.213	0.335	0.473
	0.311	0.222	0.345	0.484	0.306	0.219	0.341	0.480
MuRE	0.493	0.388	0.549	0.693	0.485	0.381	0.539	0.686
	0.267	0.191	0.289	0.420	0.270	0.194	0.292	0.424
	0.272	0.195	0.295	0.426	0.274	0.198	0.294	0.429

Table 4. Link prediction performance of different embedding models with 32-dimensional real-valued vectors on the Mutagenesis dataset without and with combined training with LitEm, Each model’s results on training, test, and validation sets are shown in three rows; bold marks the best per-metric value for each model.

Models	KGE				KGE + LitEm			
	MRR	H@1	H@3	H@10	MRR	H@1	H@3	H@10
ComplEx	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
	0.226	0.158	0.249	0.357	0.174	0.109	0.187	0.315
	0.218	0.149	0.240	0.345	0.172	0.107	0.186	0.314
DeCaL	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
	0.130	0.077	0.131	0.241	0.154	0.093	0.160	0.280
	0.128	0.073	0.134	0.244	0.152	0.089	0.160	0.280
DistMult	1.000	1.000	1.000	1.000	0.997	0.995	1.000	1.000
	0.177	0.095	0.206	0.322	0.239	0.161	0.269	0.372
	0.181	0.098	0.210	0.324	0.243	0.165	0.272	0.377
Keci	1.000	1.000	1.000	1.000	0.933	0.875	1.000	1.000
	0.103	0.059	0.105	0.174	0.158	0.109	0.162	0.246
	0.095	0.054	0.094	0.166	0.159	0.109	0.162	0.252
OMult	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
	0.070	0.030	0.061	0.149	0.202	0.102	0.283	0.389
	0.071	0.031	0.065	0.145	0.199	0.102	0.278	0.379
QMult	1.000	1.000	1.000	1.000	1.000	1.000	1.000	1.000
	0.224	0.147	0.256	0.359	0.263	0.189	0.289	0.389
	0.217	0.138	0.252	0.351	0.254	0.183	0.279	0.379
TransE	0.779	0.721	0.817	0.882	0.769	0.710	0.806	0.877
	0.563	0.473	0.620	0.713	0.568	0.483	0.624	0.704
	0.548	0.459	0.599	0.696	0.555	0.475	0.602	0.687
MuRE	1.000	0.999	1.000	1.000	1.000	1.000	1.000	1.000
	0.541	0.461	0.554	0.758	0.532	0.470	0.547	0.626
	0.518	0.435	0.533	0.742	0.511	0.450	0.526	0.602

References

1. Wang, J., Ilievski, F., Szekely, P., Yao, K.T.: Augmenting knowledge graphs for better link prediction. arXiv preprint arXiv:2203.13965 (2022)