Semantic Interpretations of Multimodal Embeddings towards Explainable Al

Nitisha Jain

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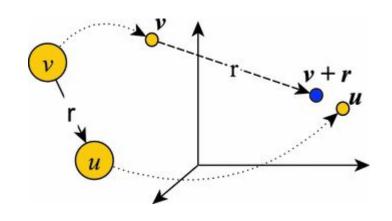




Knowledge Graph Embeddings

- Embed components of KG (entities, relations) into continuous vector spaces
- Allow easy manipulation of data while preserving inherent structure of KG
- Several popular models TransE, RESCAL, DistMult, ComplEx, ConvE ..

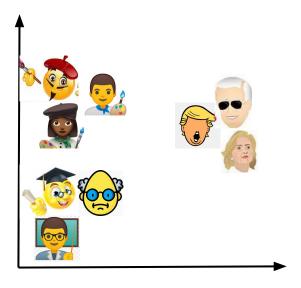
KG triple <v, r, u >



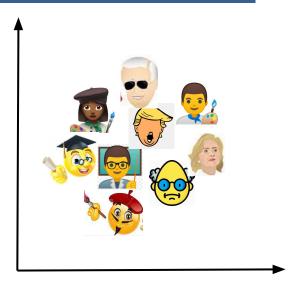
Translation based KG embedding

Expectation vs. Reality





Entity similarity reflected by vectors



Fine-grained semantics not reflected*

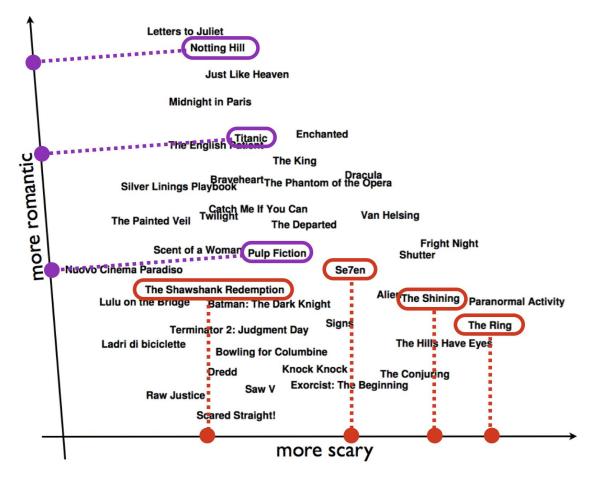
^{*}Nitisha Jain, Jan-Christoph Kalo, Wolf-Tilo Balke, Ralf Krestel: Do Embeddings Actually Capture Knowledge Graph Semantics? ESWC 2021.

Learning Quality Dimensions in Vector Spaces



- The dimensions of learned vector spaces do not normally correspond to semantically meaningful properties.
- This limits the interpretability of learned vector space representations.
- Previous work* on mitigating this issue identify interpretable directions in learned vector spaces, quality dimensions.

^{*}Joaquín Derrac and Steven Schockaert. Inducing semantic relations from conceptual spaces: A data-driven approach to plausible reasoning. Artif. Intell., 66–94, 2015.





Interpretable directions within a 2d projection of vector space embedding of movies (IMDB)

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MuseIT Use Case



- For KG embeddings of CH assets, the dimensions would correspond to the attributes of the entities.
 - E.g., Movie entities attributes could be awards, cost ...
 - Artist entities attributes could be art style, nationality ...
- Enable completion of missing attributes of entities especially helpful for multimodal representations.
- Increase semantic interpretability of the vectors explainable embeddings!

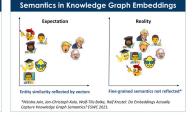
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Knowledge Graph Embeddings

- Embed components of KG (entities, relations) into continuous vector spaces.
- Allow easy manipulation of data while preserving inherent structure of KG.
- Several popular models TransE, RESCAL, DistMult, Complex, ConvE ..

Premise: Vectors of entities, relations reflect their latent semantics





Learning Interpretable Quality Dimensions in Vector Spaces

- The dimensions of learned vector spaces do not normally correspond to semantically meaningful properties.
- This limits the interpretability of learned vector space representations.
- Previous work* on mitigating this issue Identify interpretable directions in learned vector spaces, these directions can then play the role of quality dimensions.
- The central aim is to decompose the given vector space into a number of lower-dimensional spaces, each of which captures a different aspect of meaning.

*Joaquin Derrac and Steven Schockaert. Inducing semantic relations from conceptual spaces: A data-driven approach to plausible reasoning. Artif. Intell., 66–94, 2015.



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Objectives

- To investigate statistical and semantic models for multisensory representations of Cultural Heritage (CH) assets.
- To devise innovative methodologies to transform CH modalities (text, audio, video, haptics), and generate them
 automatically from existing ones with machine learning and crowdsourcing.

Applicability

- · For embeddings of CH assets, the dimensions would correspond to the attributes of the CH entities.
- o E.g., Movie entities attributes could be awards, cost .. Artist entities art style, nationality .
- · This would increase semantic interpretability of the vectors.
- Use case for MuseIT Enable completion of missing attributes of entities.
- Especially helpful for enhancing multimodal representations with embeddings



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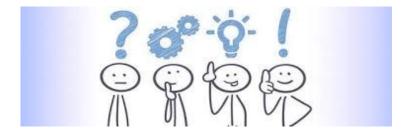






Find us at the Poster session!

We would love to discuss more!



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