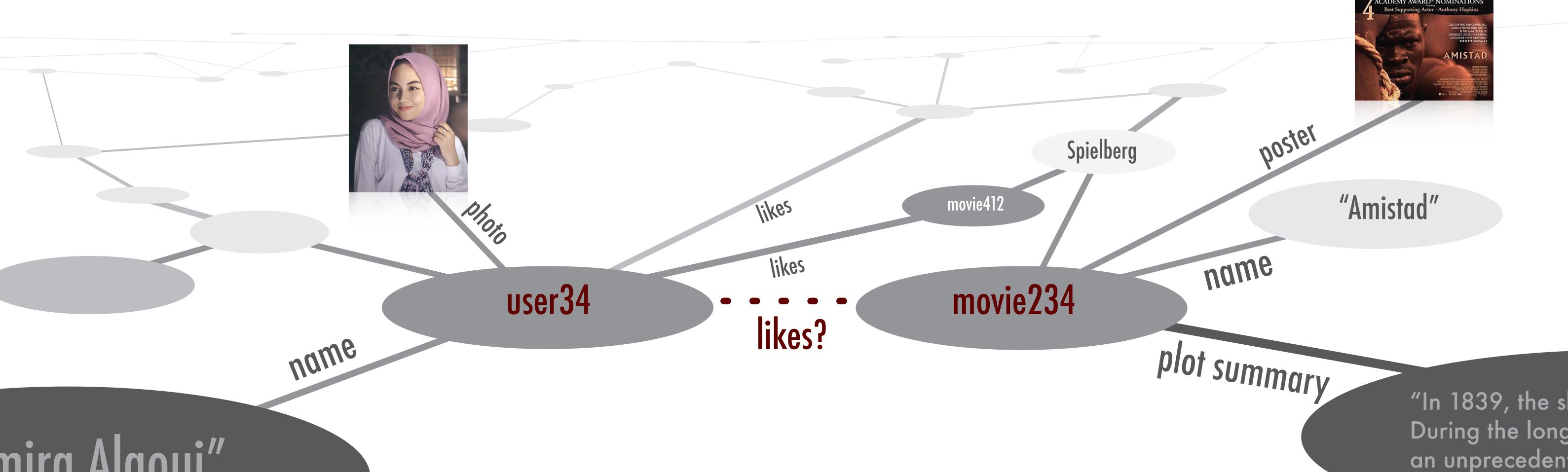
the KNOWLEDGE GRAPH as the default data model for machine learning

PETER BLOEM (VU@PETERBLOEM.NL) FRANK VAN HARMELEN PIETER ADRIAANS

The knowledge graph as the default data model for learning on heterogeneous knowledge X. Wilcke P. Bloem V. de Boer doi.org/10.3233/DS-170007 graphs natively.

End-to-end learning is crucial in complex pipelines, to combat error propagation. When input knowledge is encoded in different modalities (images, language, facts) we must design models that consume all the data as-is and convert it to the target attribute in an end-to-end fashion. To achieve this, we can represent our data as a knowledge graph, and design models that consume knowledge



Graph convolutions for heterogeneous knowledge

Graph convolutions are one of the most promising ways of processing data encoded in graphs. The *RGCN model*¹ extends this idea to knowledge graphs.

A more refined version is the *LGCN model*² which can learn edge weights adaptively.

Meaningful subunits: network motifs for knowledge graphs

How do we extract all information about user34 from the graph, when everything is connected to everything else?

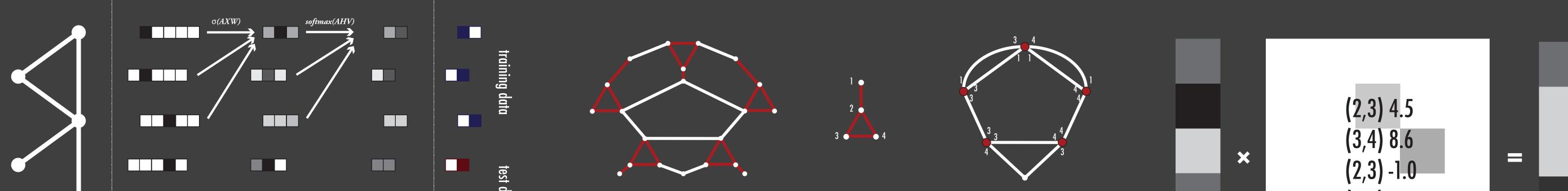
Network motifs allow us to extract the fundamental building blocks of a graph, and determine which nodes "belong together."

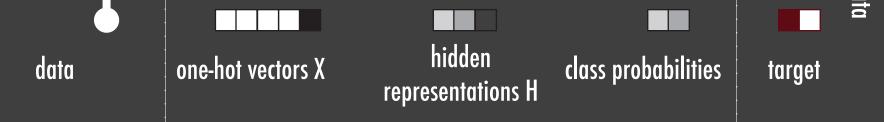
the subgraph G'

the template graph H

Sparse, adaptive hyperlayers

To learn a transformation from one graph to another, or to learn a generative model over graphs, we need a sparse transformation. Many sparse layers exist already (like convolutional layers), but the sparse structure must be known in advance. When the sparsity is informed by the data, and changes from one instance to the next, a new approach is required.





¹Modeling Relational Data with Graph Convolutional Networks arxiv.org/abs/1703.06103

²End-to-end learning of latent edge weights for Graph **Convolutional Networks** (MSc thesis) esc.fnwi.uva.nl/thesis/apart/ki

Finding Network Motifs in Large Graphs using Compression as a Measure of Relevance arxiv.org/abs/1701.02026

Detecting Motifs in Knowledge Graphs using Compression github.com/MaestroGraph/motive-rdf

the data G

(5,5) 0.5 input x sparse weight matrix V output y hypernetwork f(x)

Sparse adaptive, hyperlayers github.com/MaestroGraph/sparse-hyper



AAA Data Science project: The structure of knowledge