Embeddings:
Vector & Semantic Search

Applications | Concepts | Examples

January 11 2024 | Lecture at University of Washington, Seattle by Dr. Karthik Mohan

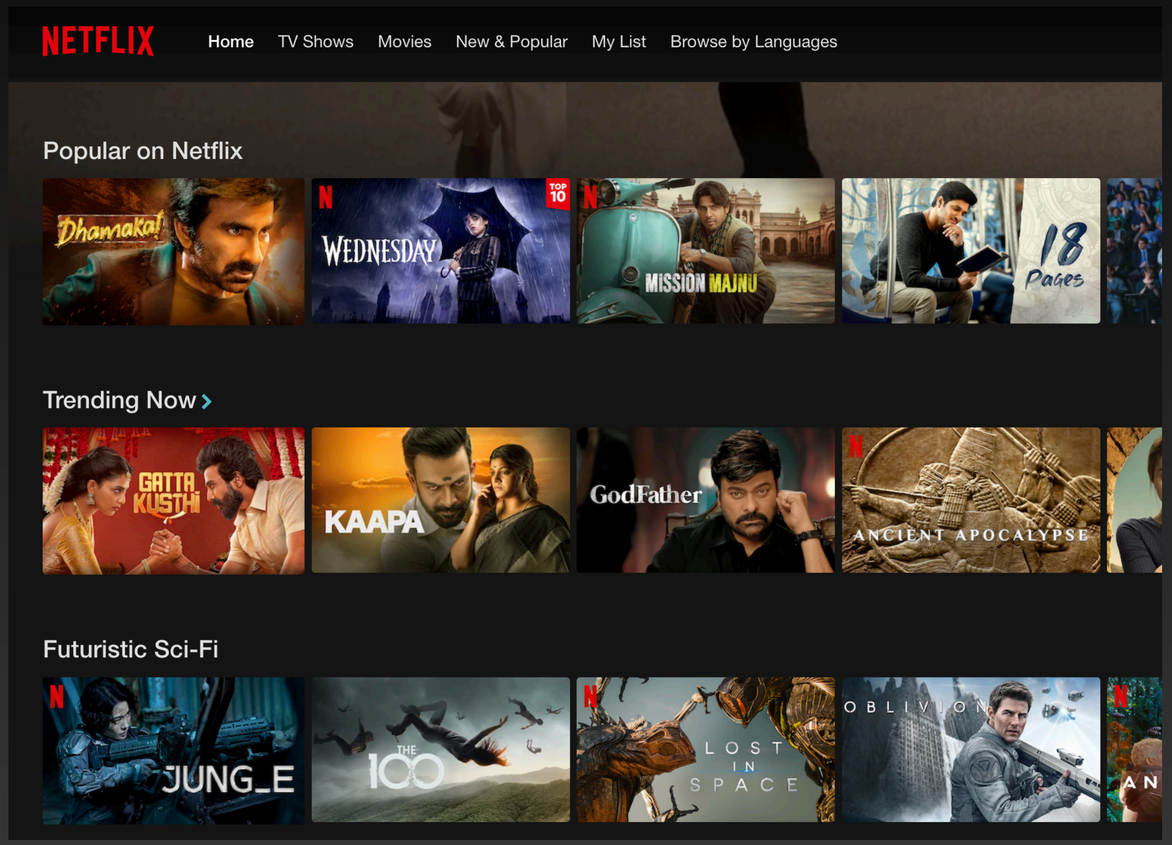
Today's Talk

1. Introduction and Motivation

2. Semantic Search

Netflix Recommendations

MULTIPLE CAROUSELS

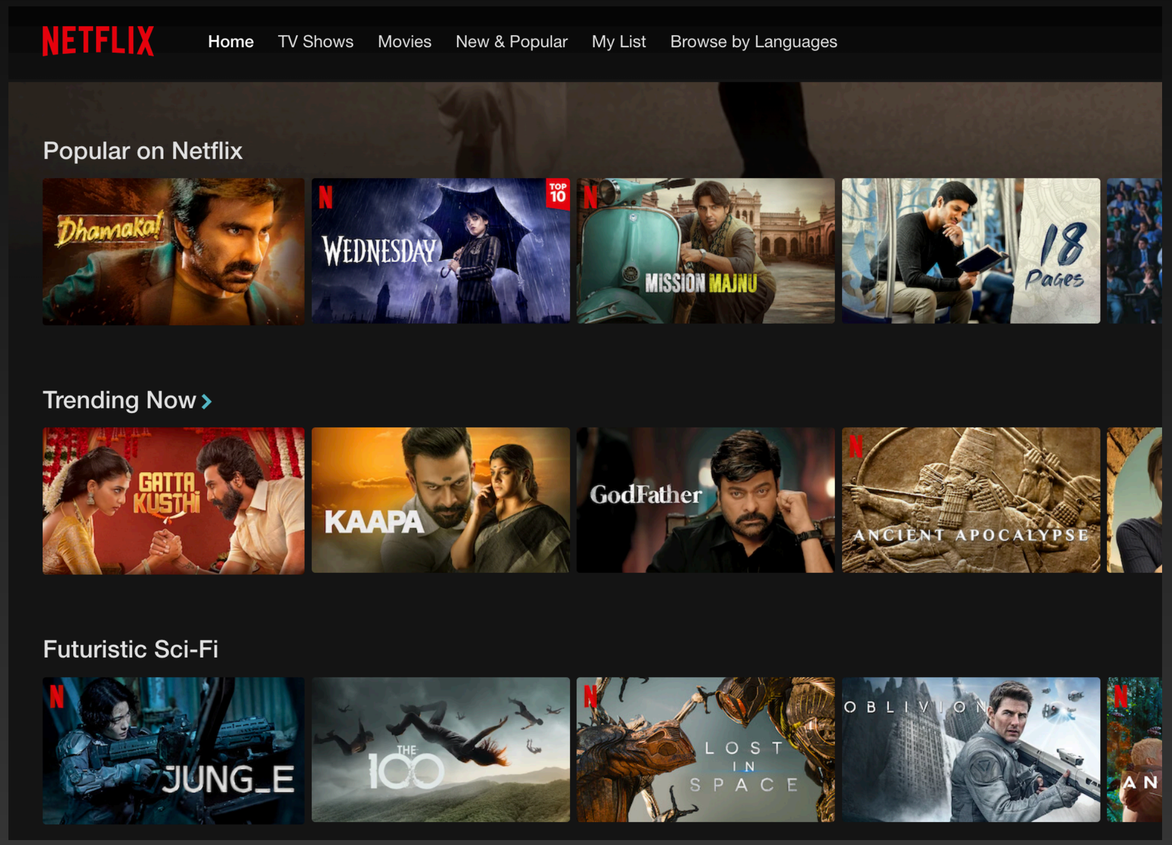


Netflix Recommendations

POPULAR yet PERSONALIZED

PERSONALIZED TRENDS

PERSONALIZED GENRE



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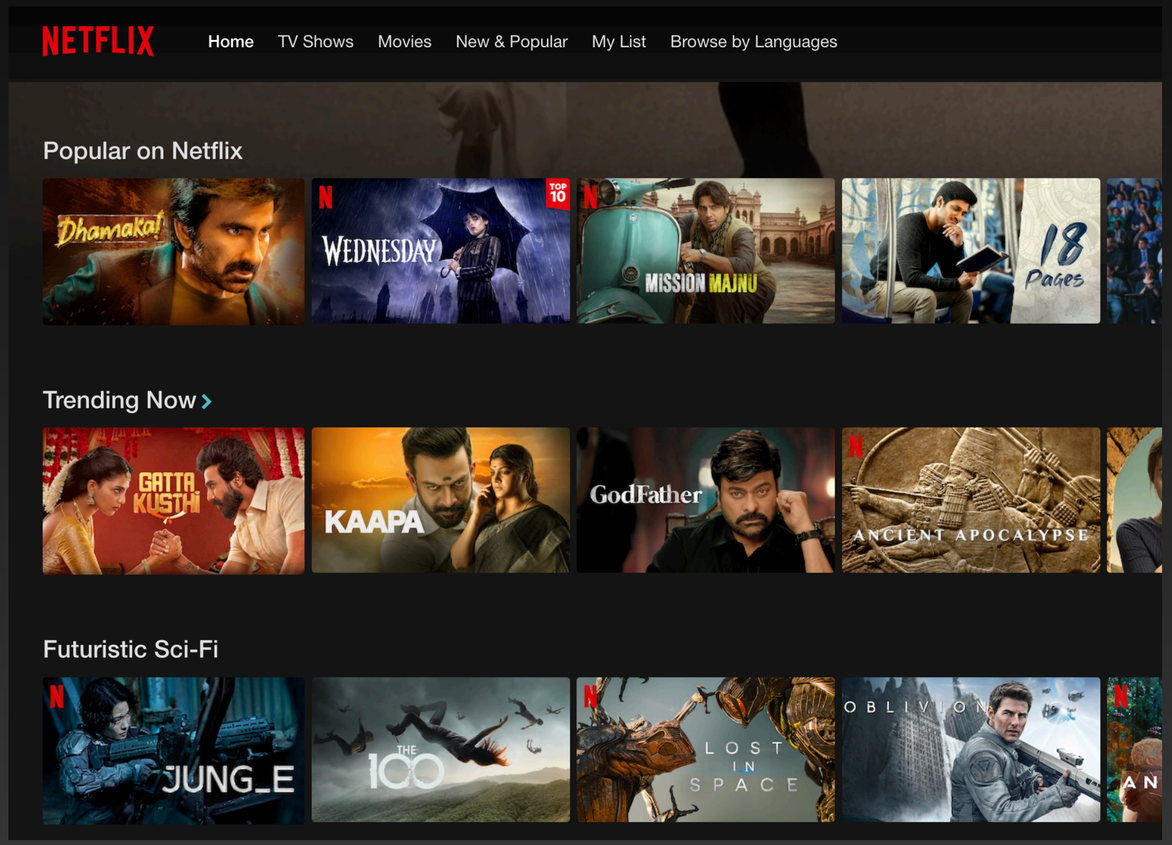

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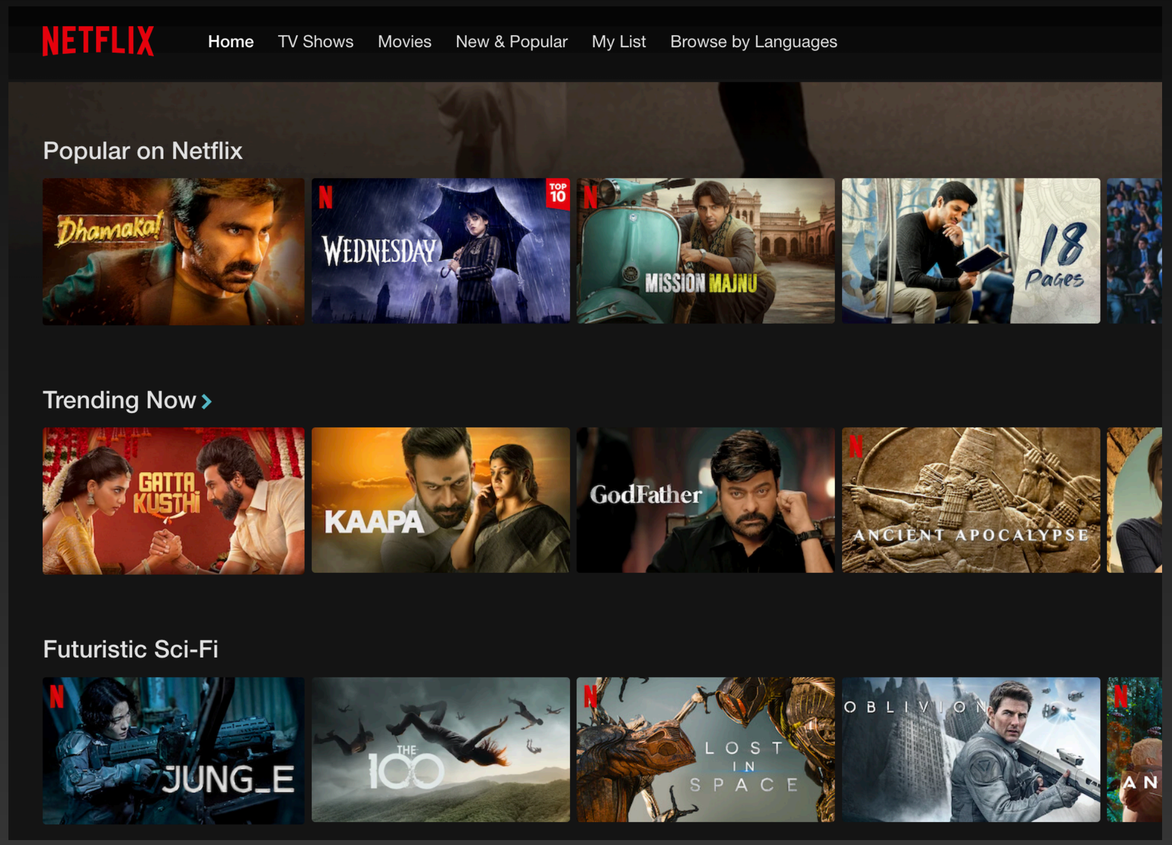
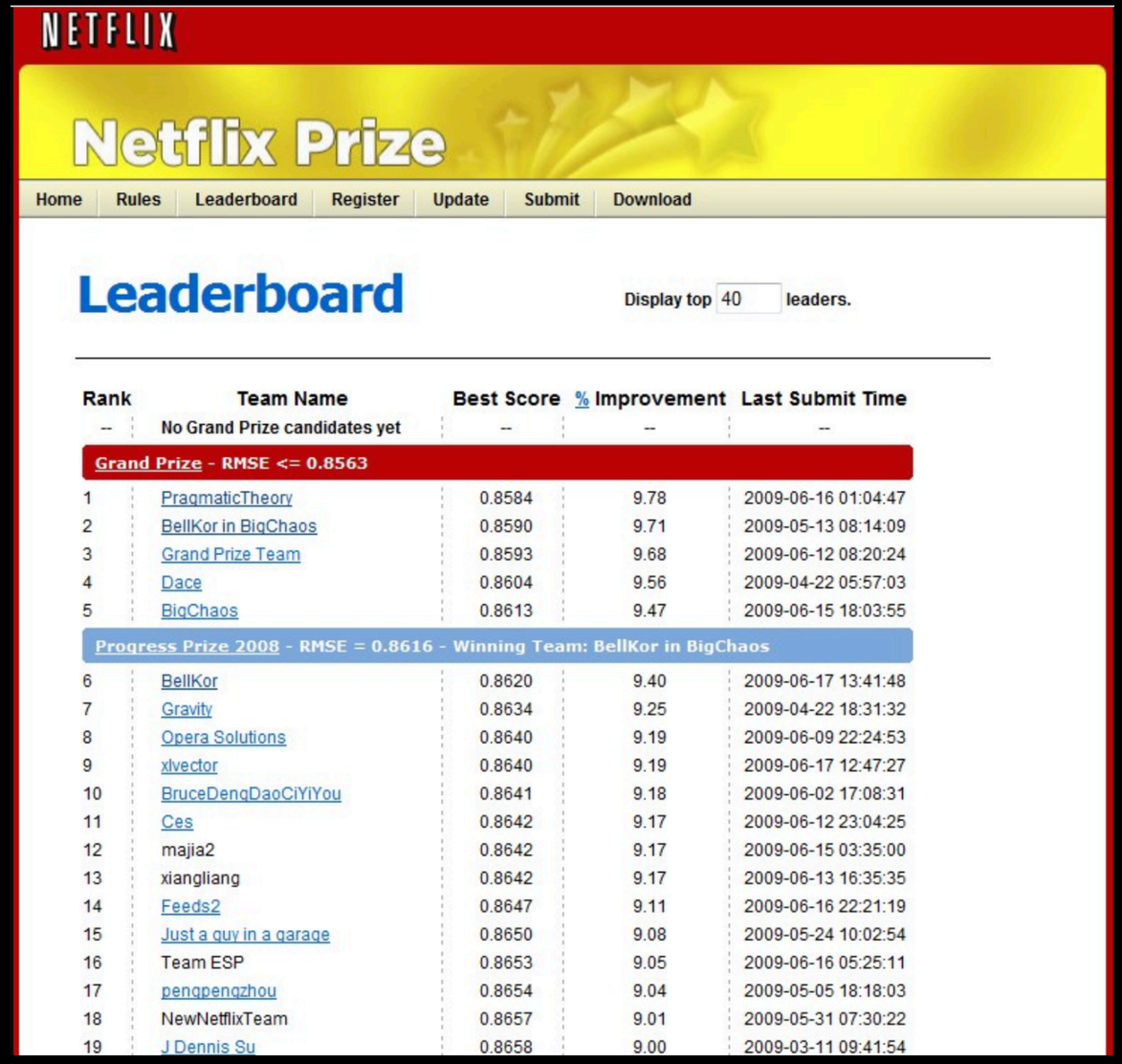


Netflix Million Dollar Prize!

Netflix Million Dollar Prize!



NETFLIX

Netflix Prize

Home | Rules | Leaderboard | Register | Update | Submit | Download

Leaderboard

Display top 40 leaders.

Rank	Team Name	Best Score	% Improvement	Last Submit Time
--	No Grand Prize candidates yet	--	--	--
Grand Prize - RMSE <= 0.8563				
1	PragmaticTheory	0.8584	9.78	2009-06-16 01:04:47
2	BellKor in BigChaos	0.8590	9.71	2009-05-13 08:14:09
3	Grand Prize Team	0.8593	9.68	2009-06-12 08:20:24
4	Dace	0.8604	9.56	2009-04-22 05:57:03
5	BigChaos	0.8613	9.47	2009-06-15 18:03:55
Progress Prize 2008 - RMSE = 0.8616 - Winning Team: BellKor in BigChaos				
6	BellKor	0.8620	9.40	2009-06-17 13:41:48
7	Gravity	0.8634	9.25	2009-04-22 18:31:32
8	Opera Solutions	0.8640	9.19	2009-06-09 22:24:53
9	xlvector	0.8640	9.19	2009-06-17 12:47:27
10	BruceDengDaoCiYiYou	0.8641	9.18	2009-06-02 17:08:31
11	Ces	0.8642	9.17	2009-06-12 23:04:25
12	majia2	0.8642	9.17	2009-06-15 03:35:00
13	xiangliang	0.8642	9.17	2009-06-13 16:35:35
14	Feeds2	0.8647	9.11	2009-06-16 22:21:19
15	Just a guy in a garage	0.8650	9.08	2009-05-24 10:02:54
16	Team ESP	0.8653	9.05	2009-06-16 05:25:11
17	pengpengzhou	0.8654	9.04	2009-05-05 18:18:03
18	NewNetflixTeam	0.8657	9.01	2009-05-31 07:30:22
19	J Dennis Su	0.8658	9.00	2009-03-11 09:41:54

Collaborative Filtering

Rishi

Michael

Karthik

Roshin

Amy

Collaborative Filtering

Avatar

Arrival

When Harry

Before Sunrise

Minions

Rishi

Michael

Karthik

Roshin

Amy

Collaborative Filtering

	 Avatar	 Arrival	 When Harry	 Before Sunrise	 Minions
Rishi	👍	👍	👎	👎	👍
Michael	👎		👍	👍	👍
Karthik	👍		👎		👍
Roshin		👎	👍	👍	
Amy		👍		👍	👎

Collaborative Filtering

	 Avatar	 Arrival	 When Harry	 Before Sunrise	 Minions
Rishi	👍	👍	👎	👎	👍
Michael	👎	❓	👍	👍	👍
Karthik	👍	❓	👎	❓	👍
Roshin	❓	👎	👍	👍	❓
Amy	❓	👍	❓	👍	👎

Collaborative Filtering

	 Avatar	 Arrival	 When Harry	 Before Sunrise	 Minions
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Karthik	👍	❓	👎	❓	👍
Roshin	❓	👎	👍	👍	❓
Amy	❓	👍	❓	👍	👎

Collaborative Filtering


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Amy	❓	👍	❓	👍	👎

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Amy	❓	👍	❓	👍	👎

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Amy	❓	👍	❓	👍	👎

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Roshin	👎	👎	👍	👍	👍
Amy	❓	👍	👍	👍	👎

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Karthik	👍	👍	👎	👎	👍
Roshin	👎	👎	👍	👍	👍
Amy	?	👍	👍	👍	👎

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Roshin	👎	👎	👍	👍	👍
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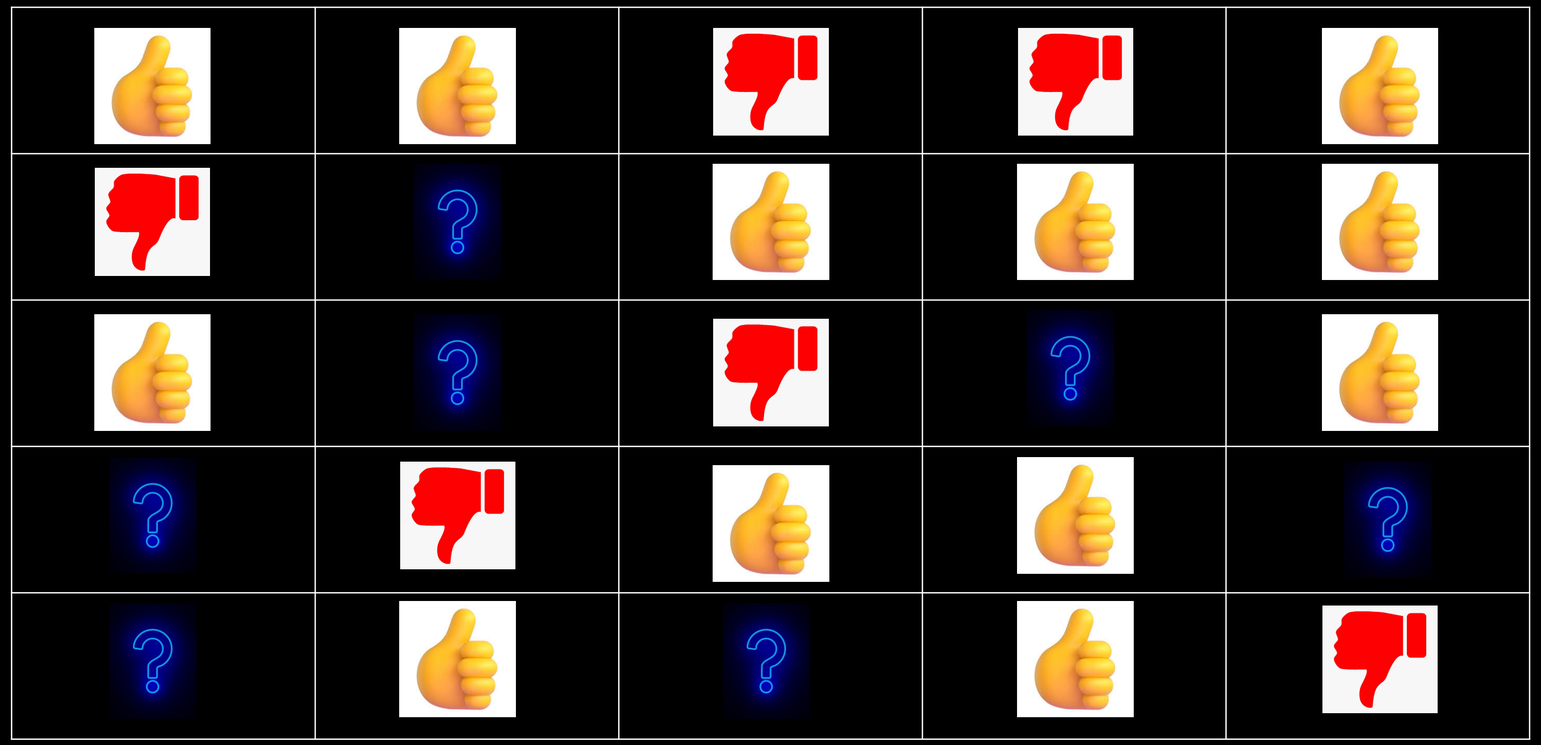
Roshin	👎	👎	👍	👍	👍
Amy	👍	👍	👍	👍	👎

Collaborative Filtering

Through Matrix Completion!



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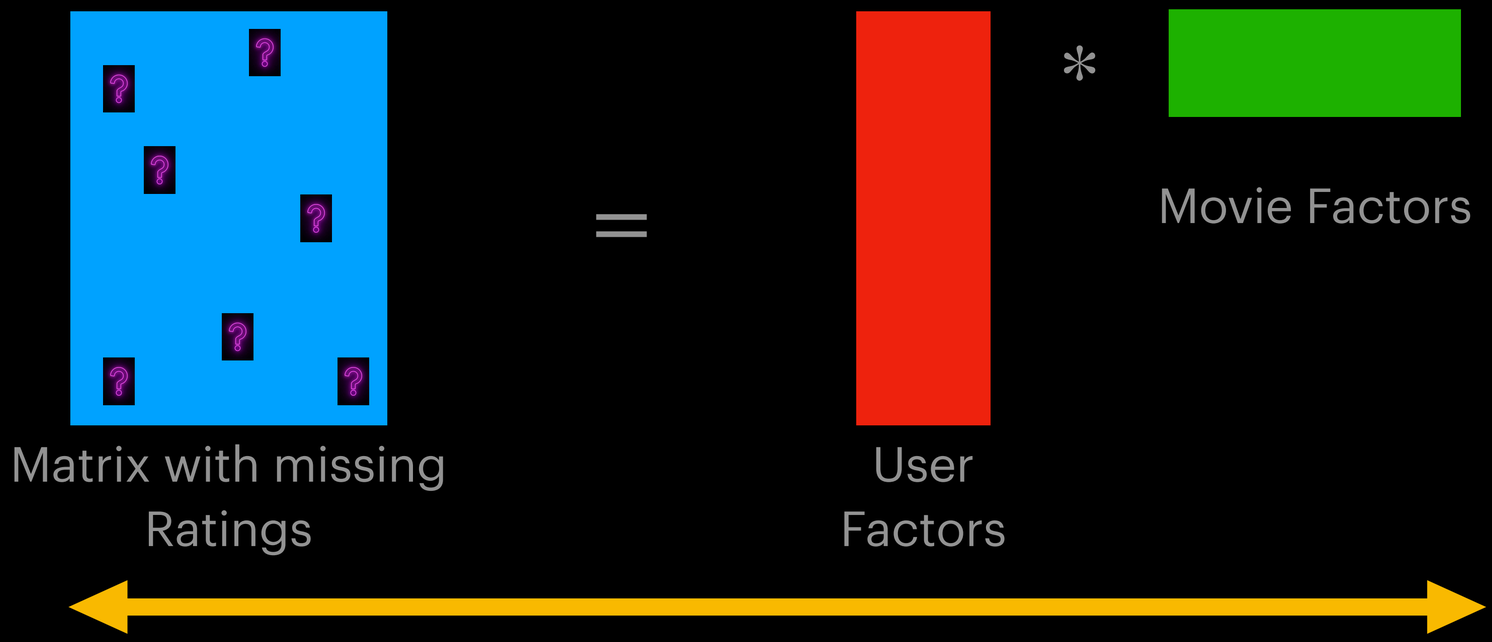

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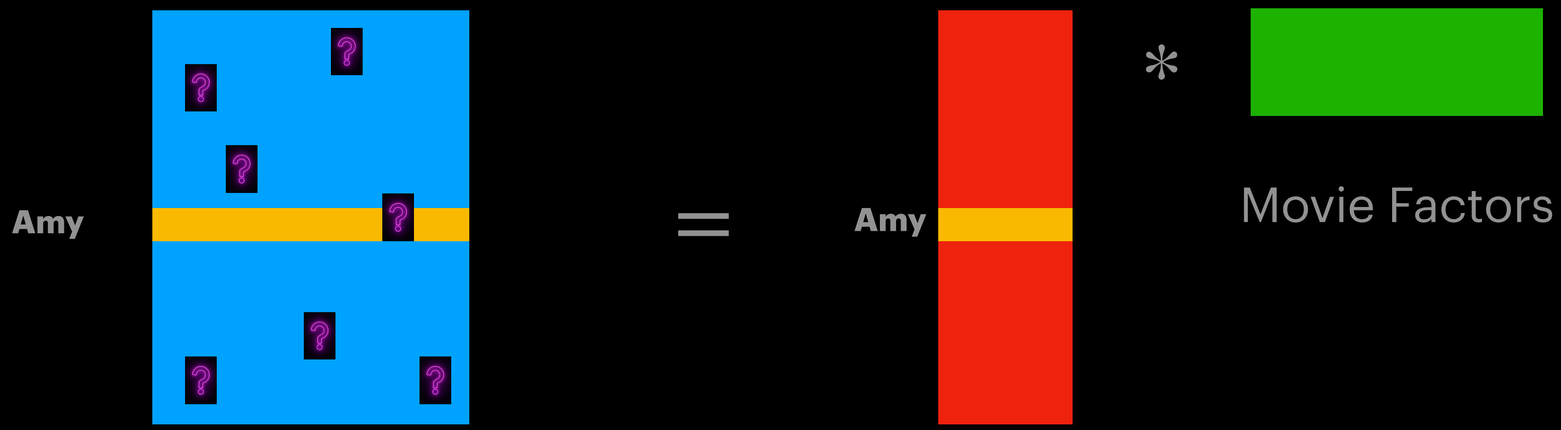

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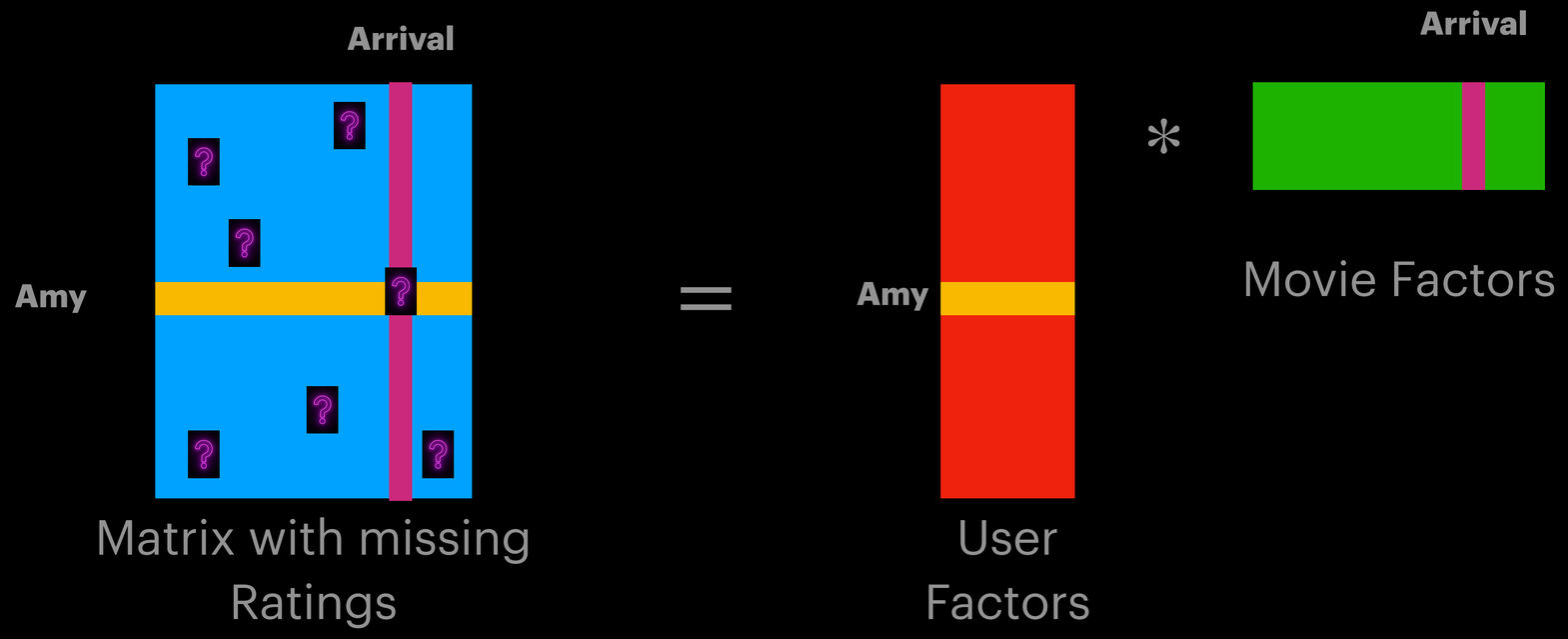

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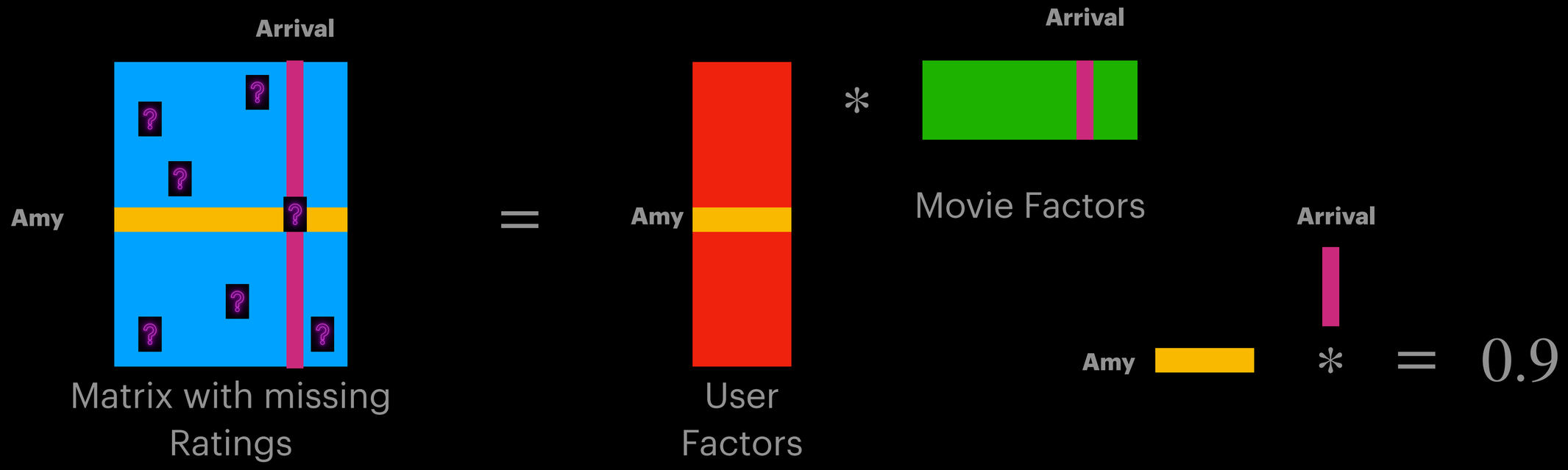

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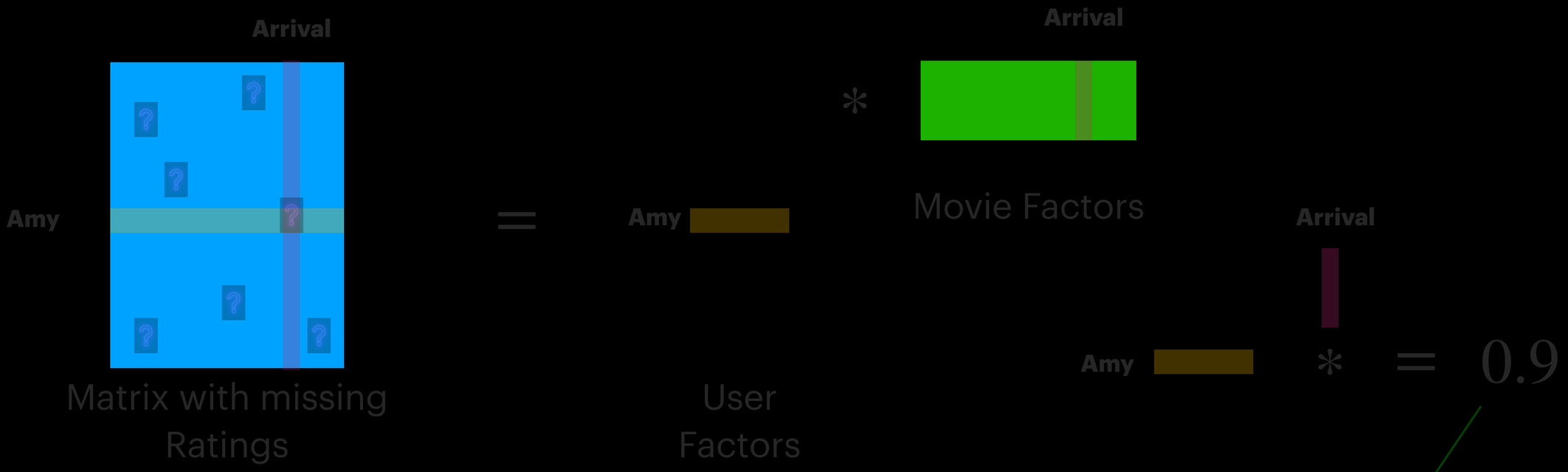


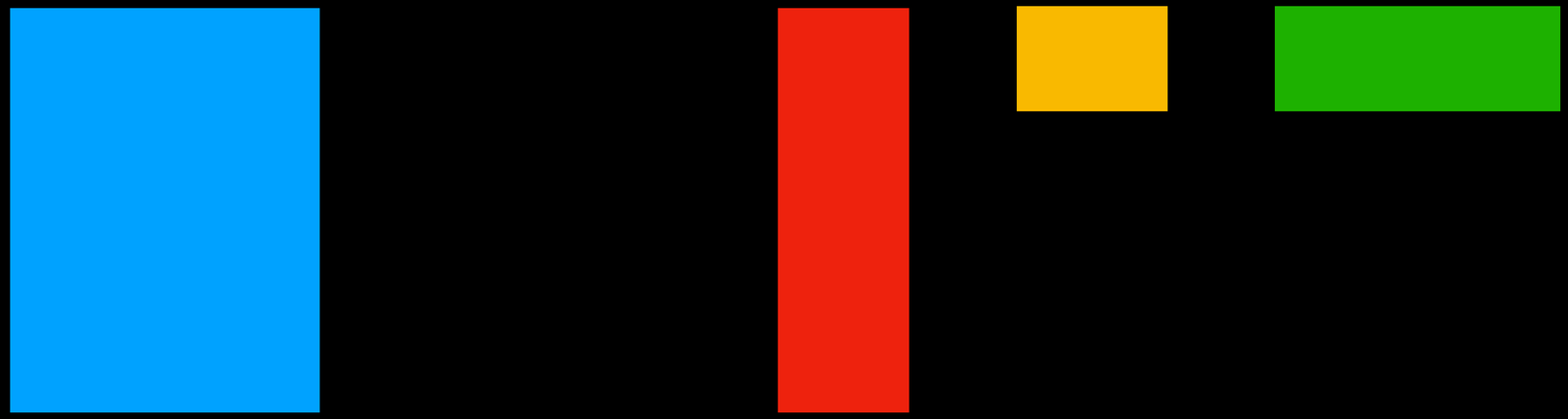
Linear Algebra in Action

Q: What is the closest Linear Algebra method that looks similar to the above factorization?

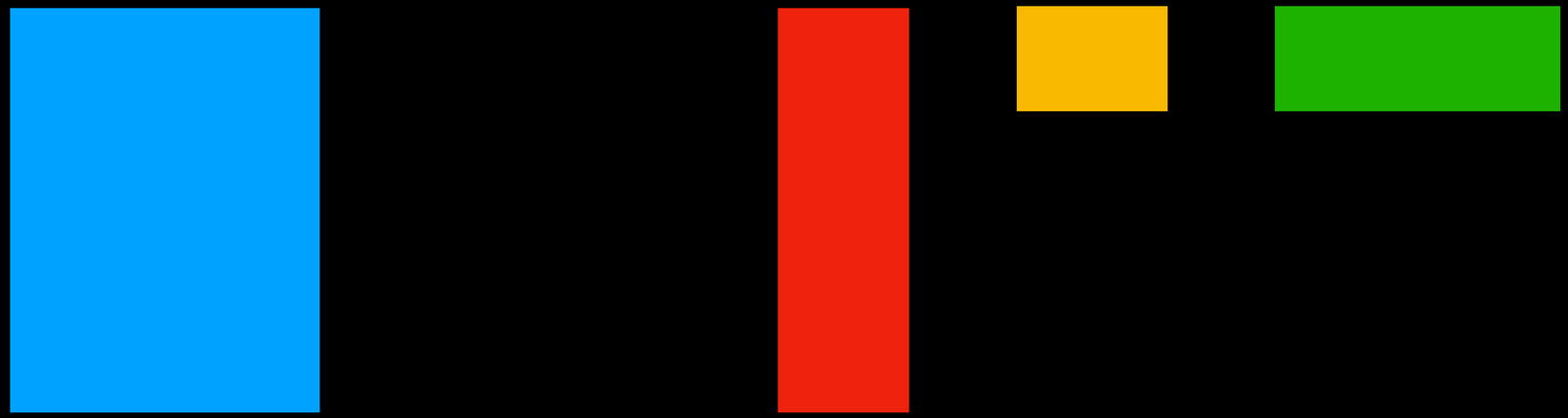
A: SVD = Singular Value Decomposition

Model predicts Amy Likes Arrival

SVD of a matrix



SVD of a matrix



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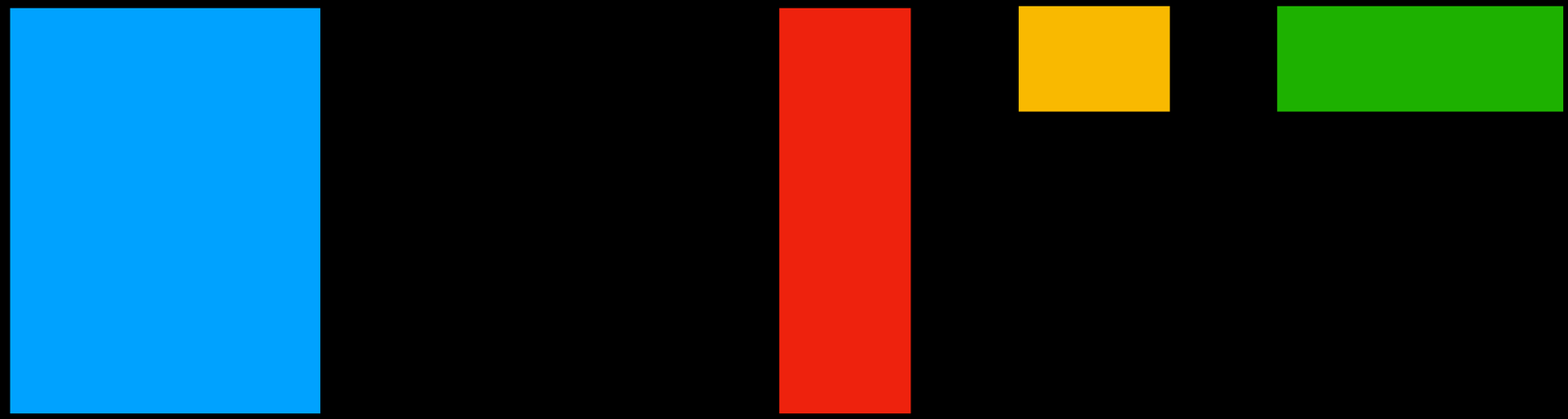
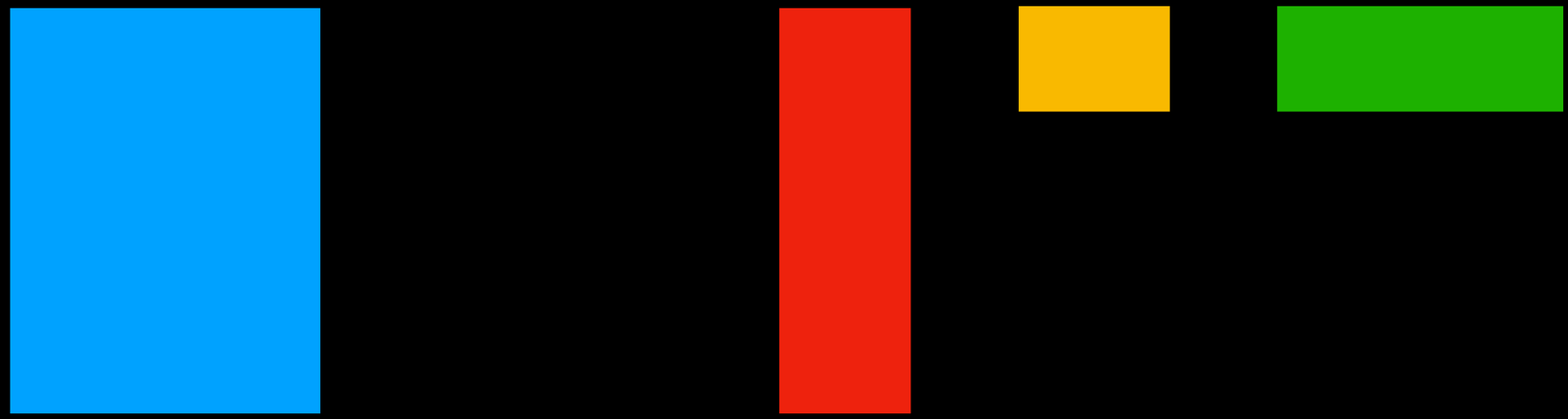


X U Sigma V^T

$$X = U\Sigma V^T$$

SVD of a matrix

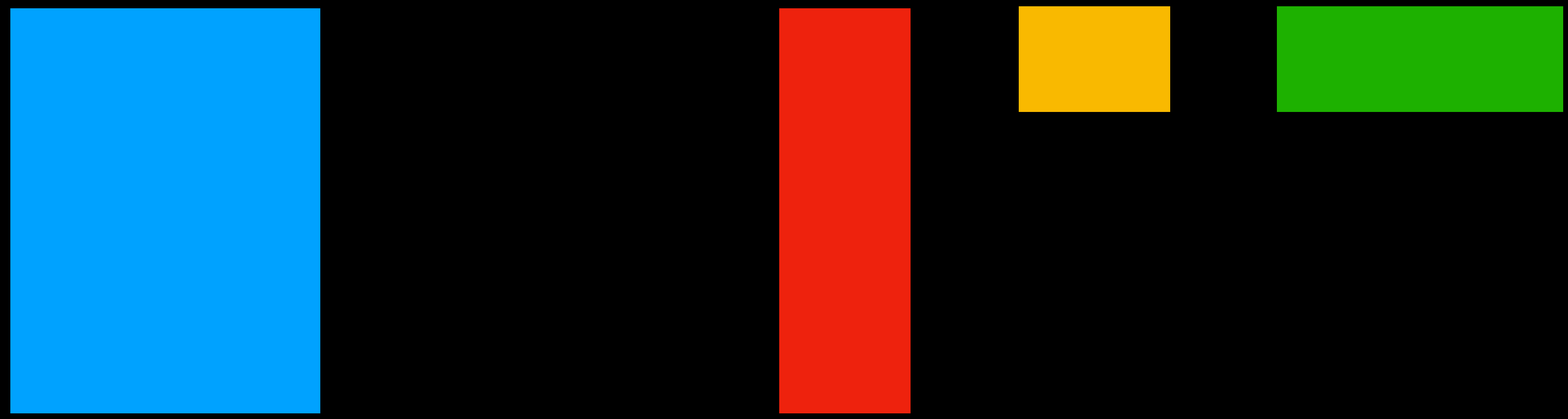
Every matrix has an SVD!



$$X = U\Sigma V^T$$

SVD of a matrix

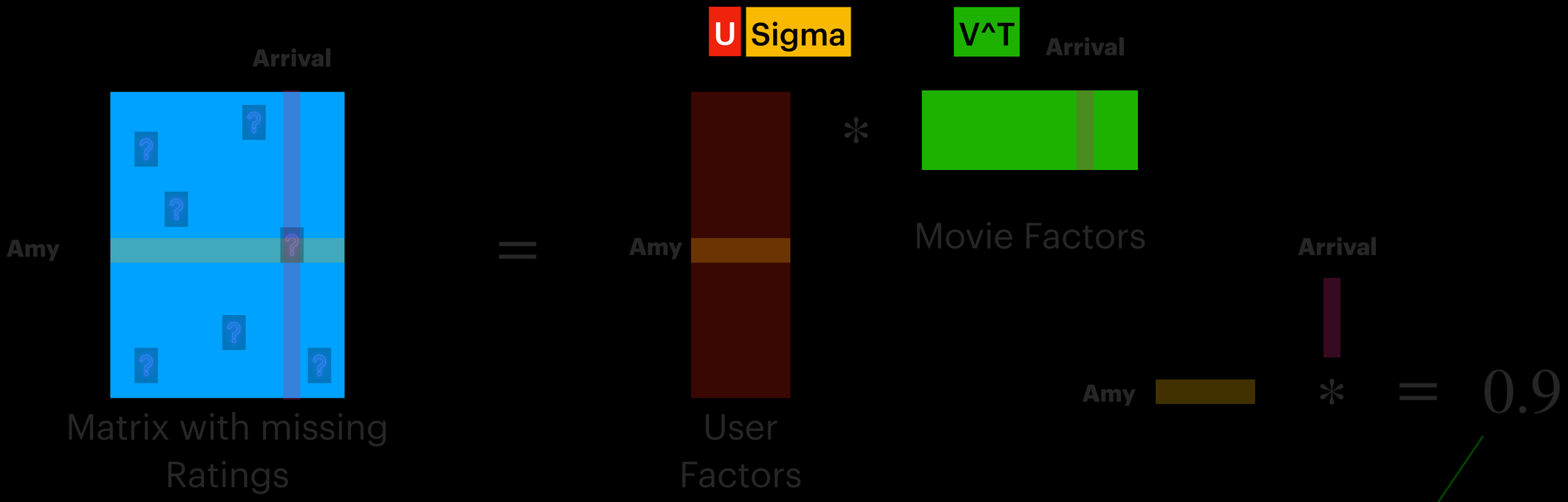
Hence: Data Matrix also has an SVD!



$$X = U\Sigma V^T$$

Collaborative Filtering

Through Matrix Completion!



Collaborative Filtering: Advanced SVD method or Iterative SVD method

Collaborative Filtering

Avatar

Arrival

When Harry

Before Sunrise

Minions

Men in Black

	Avatar	Arrival	When Harry	Before Sunrise	Minions
U1	👍	👍	👎	👎	👍
U2	👎	👎	👍	👍	👍
U3	👍	👍	👎	👎	👍
U4	👎	👎	👍	👍	👍
U5	👍	👍	👍	👍	👎

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U4	👎	👎	👍	👍	👍	?
U5	👍	👍	👍	👍	👎	?

Cold Start Problem

Content Based Filtering

vs Collaborative Filtering

Avatar

Arrival

When Harry

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Minions

Men in Black

	Avatar	Arrival	When Harry	Before Sunrise	Minions
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Karthik

Content Based Filtering

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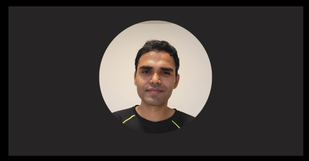

Karthik

Content Based Filtering

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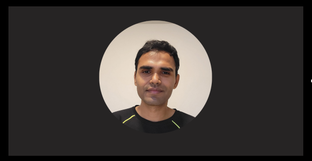

Karthik

Watched

Arrival

Content Based Filtering

vs Collaborative Filtering

Avatar

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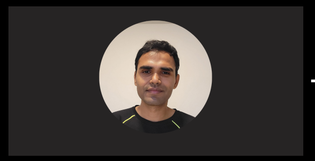

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vs Collaborative Filtering

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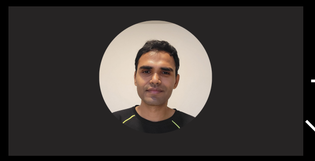

Karthik

Watched

Arrival

Likely
To
Watch

Men in Black

Content Based Filtering

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Men in Black

Men in Black

Arrival

Arrival

When Harry met

When Harry met

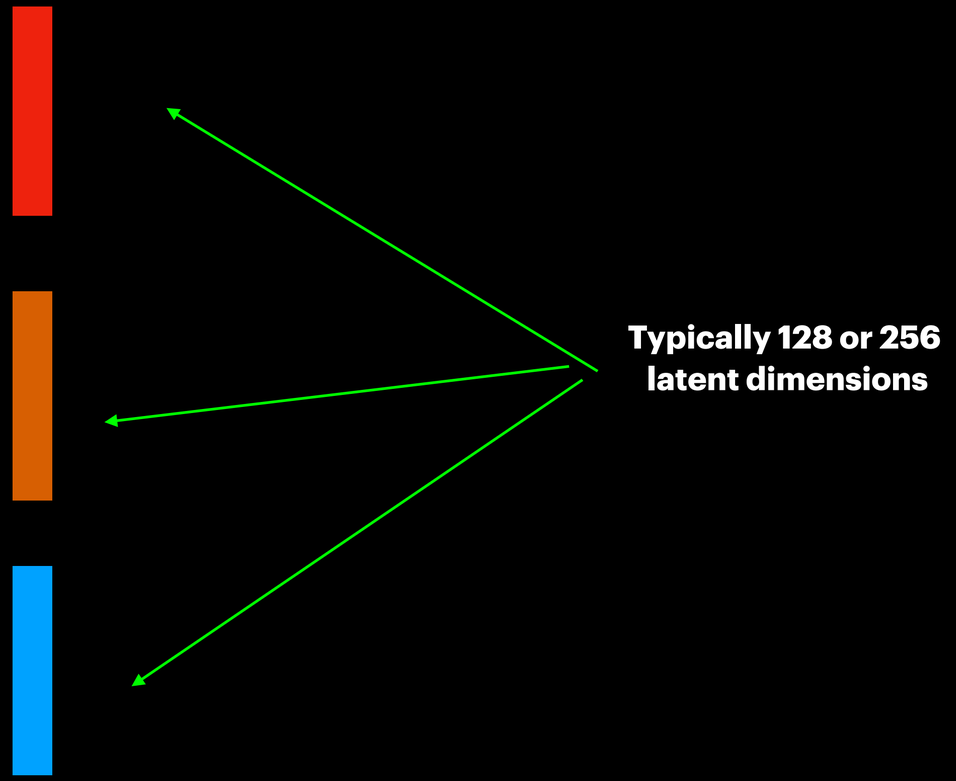
Content Based Filtering

Men in Black

Arrival

When Harry met

Embeddings



Typically 128 or 256 latent dimensions

Content Based Filtering

Embeddings

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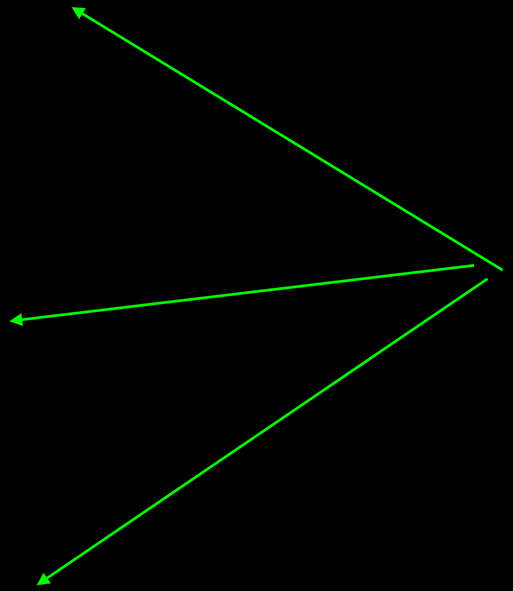

How do we
Obtain these embeddings?

Content Based Filtering

Embeddings

Men in Black

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How do we
Obtain these embeddings?

A: Through a DL model!
Maybe last but one hidden layer activations

Embeddings Interpretation

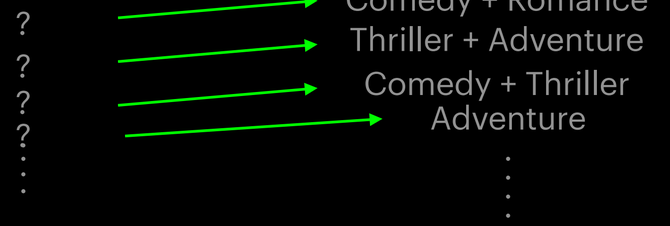
Men in Black

Arrival

When Harry met

Embeddings

Latent Dimensions → **Interpretation**



Embeddings | Vector Representations

Men in Black

Arrival

When Harry met

Embeddings

Latent Dimensions → **Interpretation**

? ? ? ? ⋮

- Comedy + Romance
- Thriller + Adventure
- Comedy + Thriller
- Adventure

⋮

Embeddings in Action

Minions

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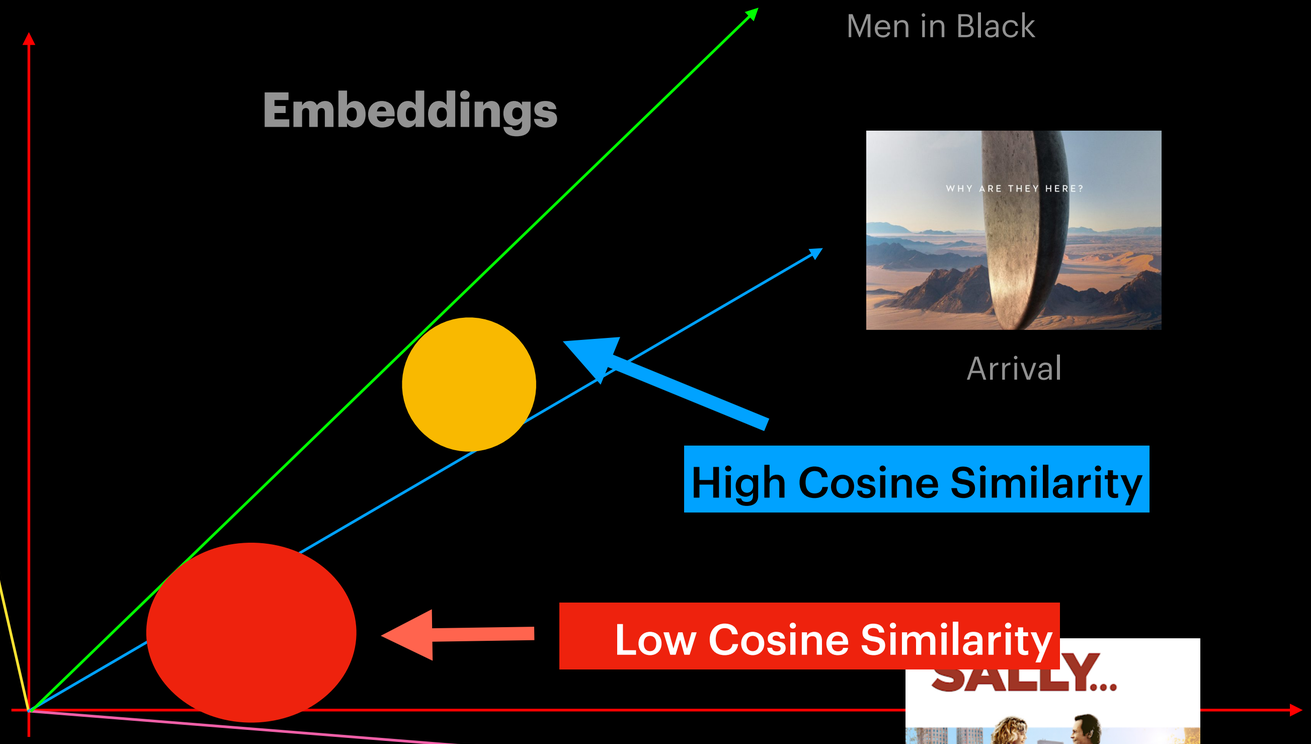


Arrival

When Harry met

Embeddings in Action

Minions

Men in Black

Men in Black

Arrival

When Harry met

Smaller Angle = Higher Cosine Similarity
Larger Angle = Lower Cosine Similarity

Arrival

When Harry met

Embeddings in Action

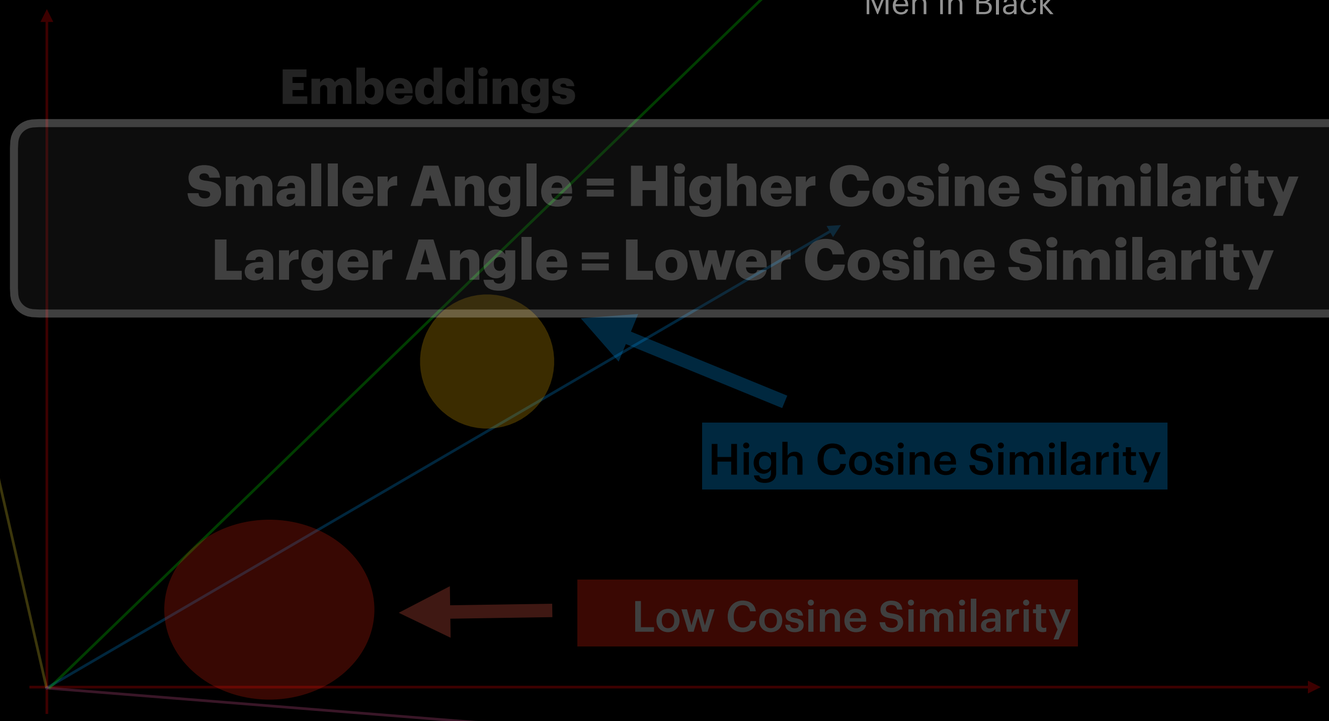
Minions

Men in Black

Men in Black

Arrival

When Harry met



$$? \leq \quad CosineSimilarity(x, y) = \frac{x^T y}{||x|| ||y||} \quad \leq ?$$

Embeddings in Action

Minions

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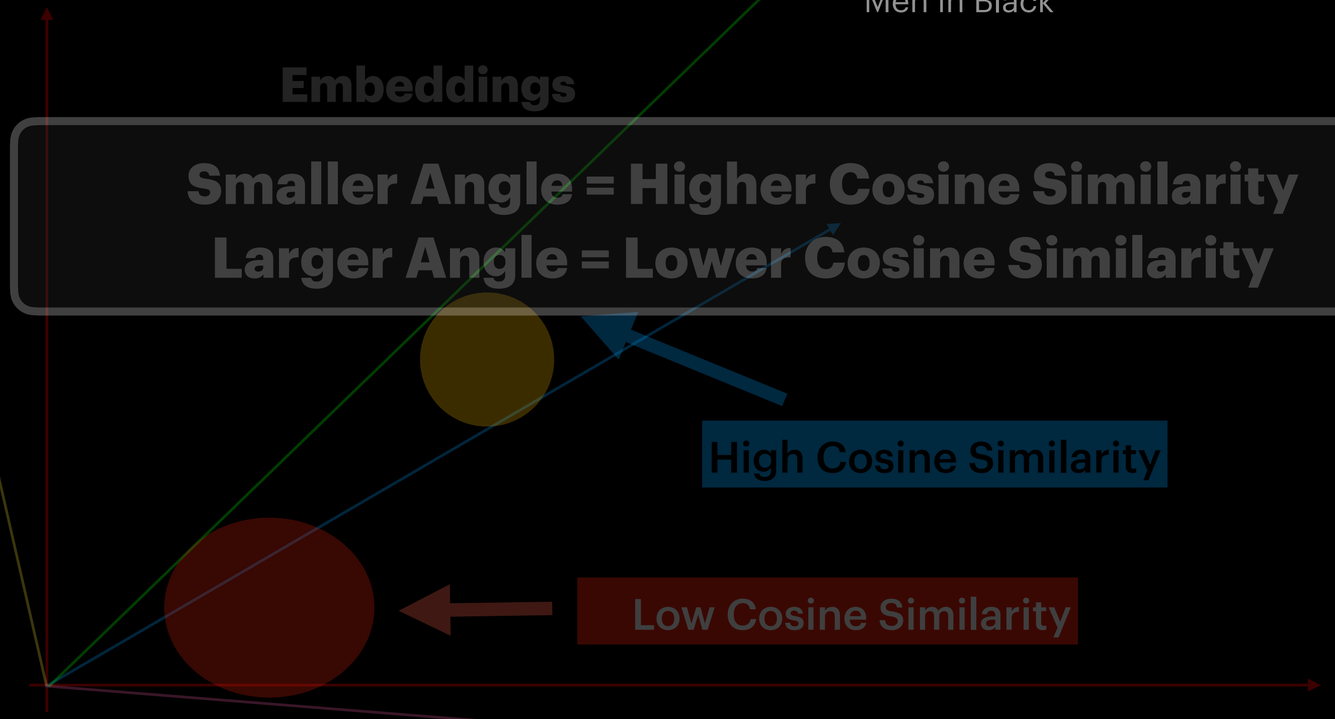


$$-1 \leq CosineSimilarity(x, y) = \frac{x^T y}{||x|| ||y||} \leq 1$$

Cauchy-Schwarz Inequality!

$$-1 \leq \ CosineSimilarity(x, y) = \frac{x^T y}{||x|| ||y||} \leq 1$$

Cauchy-Schwarz Inequality!

$$-1 \leq \ CosineSimilarity(x, y) = \frac{x^T y}{||x|| ||y||} \leq 1$$

$$|x^T y| <= ||x|| ||y||$$

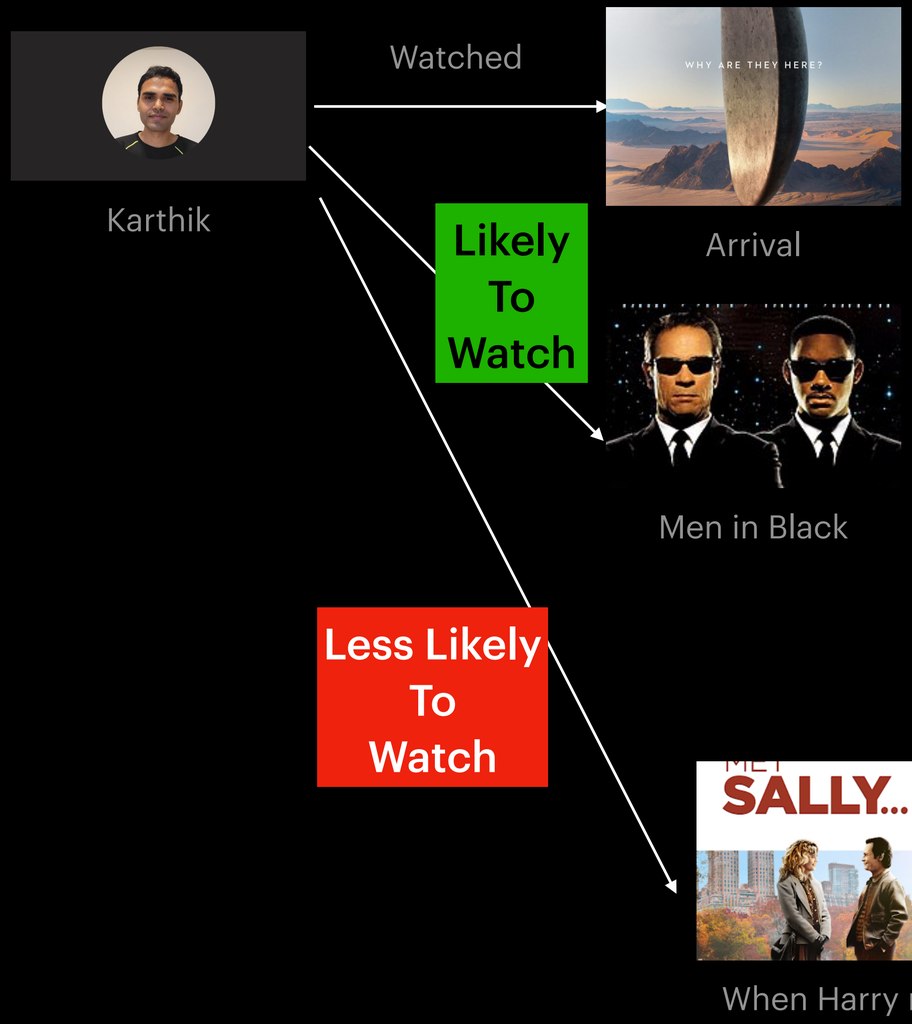
Cauchy-Schwarz Inequality!

$$-1 \leq \ CosineSimilarity(x, y) = \frac{x^T y}{||x|| ||y||} \leq 1$$

$$|x^T y| <= ||x|| ||y||$$

$||x||_2$ Euclidean Norm of x

Embeddings in Action



What if I like both sci-fi and romance?

Content Based Filtering

Minions

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When Harry met

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Content Based Filtering

Minions

Men in Black

Men in Black

Arrival

Embeddings

Arrival

When Harry met

Karthik

Can Embed both Movies and Users in Same space!

When Harry met

Content Based Filtering

Minions

Men in Black

Arrival

Karthik

When Harry met

Men in Black

Arrival

When Harry met

Content Based Filtering

Minions

Men in Black

Arrival

Embeddings

Karthik

Equal Cosine Similarity

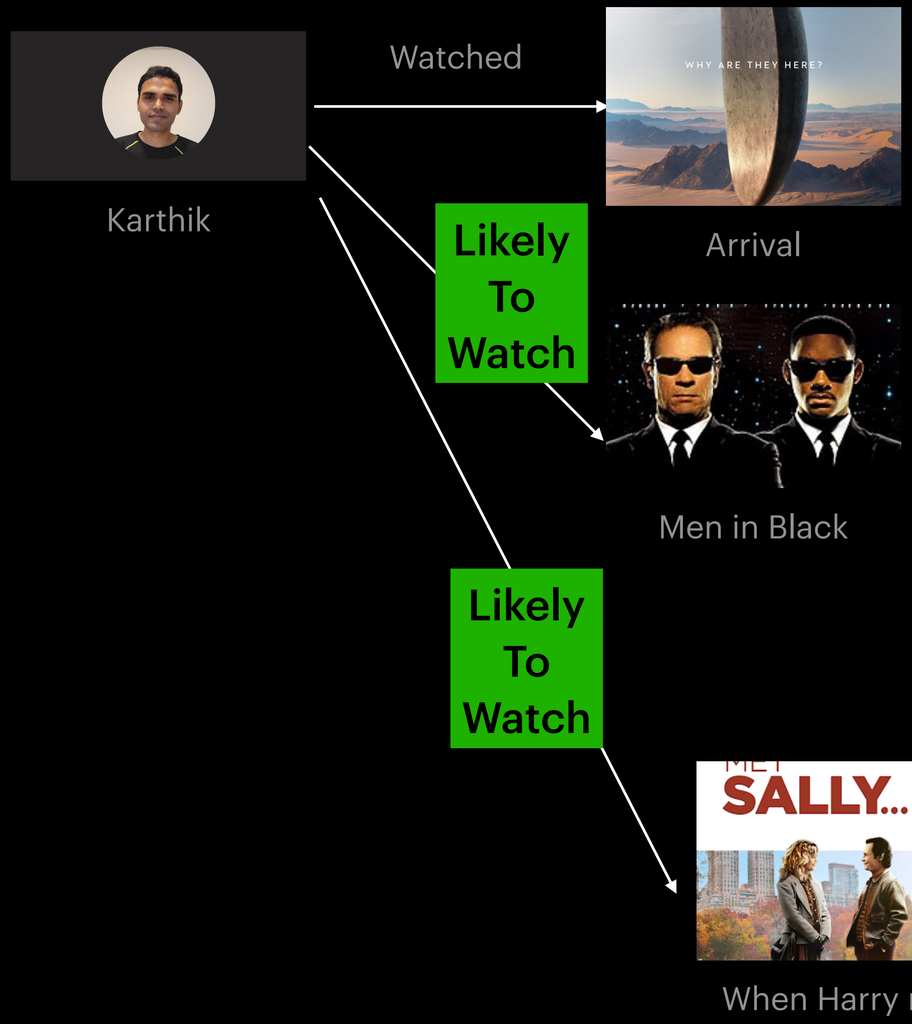
When Harry met

Men in Black

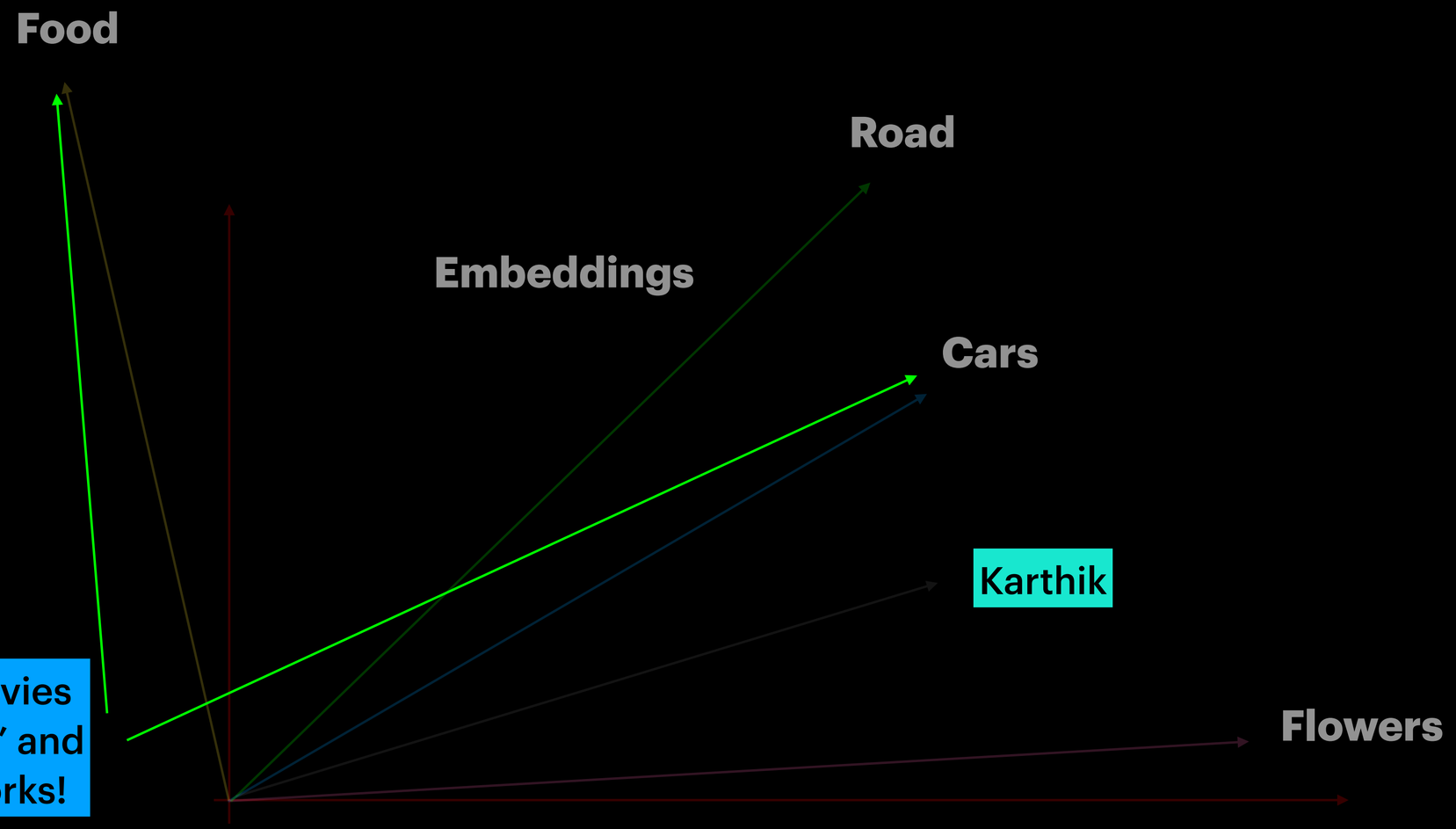
Arrival

When Harry met

Like both Sci-fi and Romance

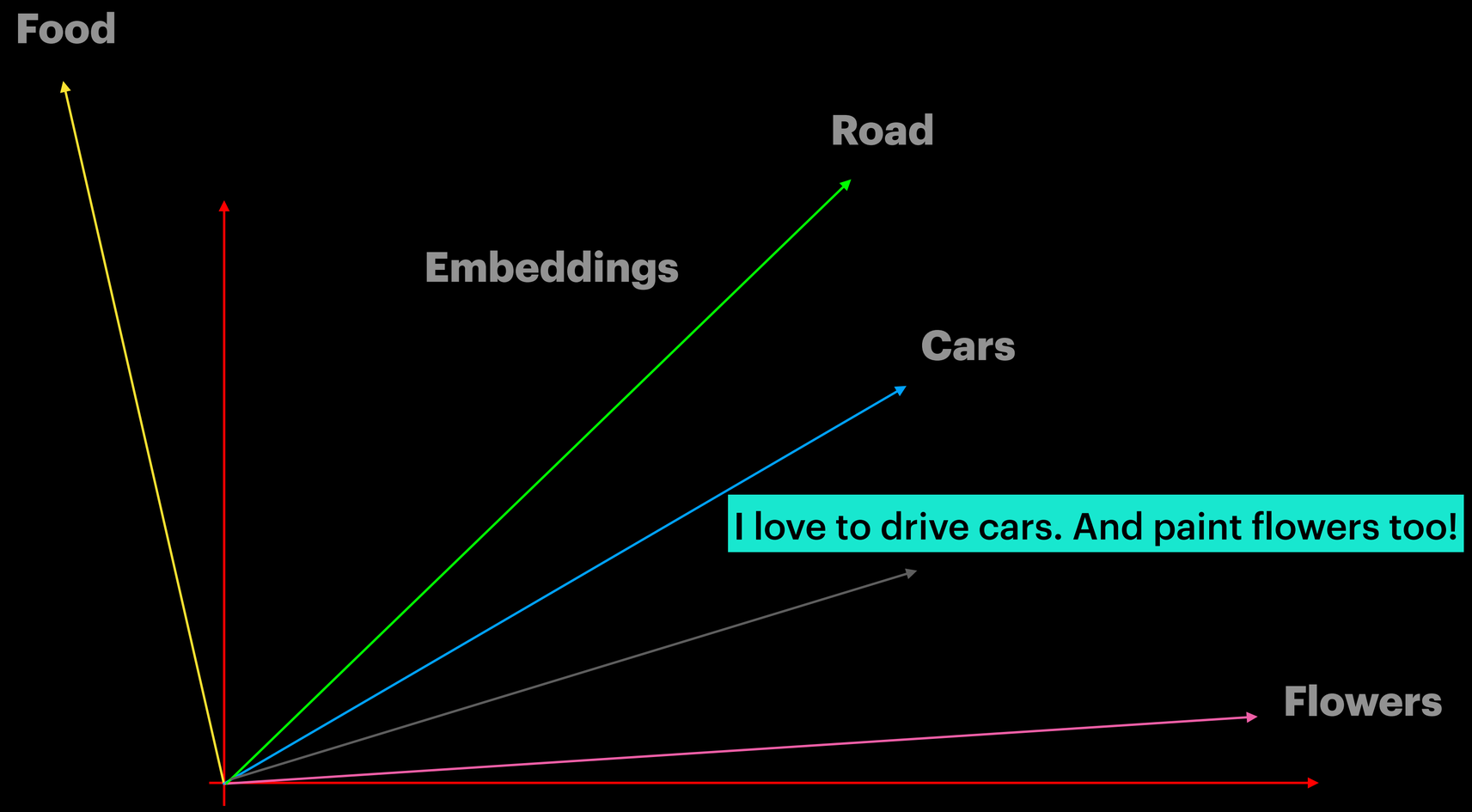


Word Embeddings

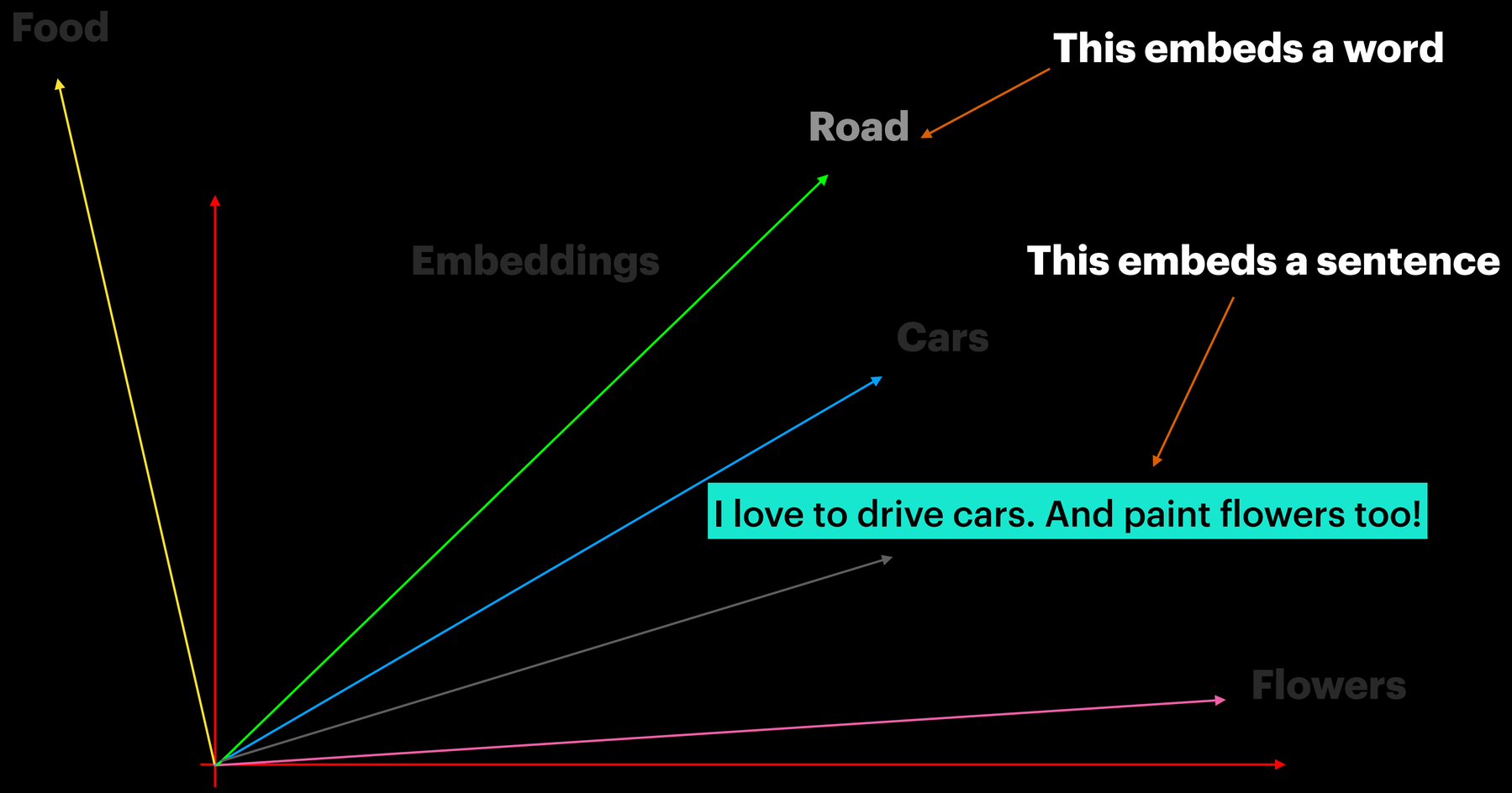


Replace movies with “words” and This still works!

Word Embeddings



Word and Sentence Embeddings



Word and Sentence Embeddings

How do we
Obtain these embeddings?

A: Through a DL model!
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Word and Sentence Embeddings

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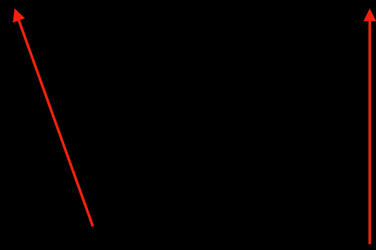
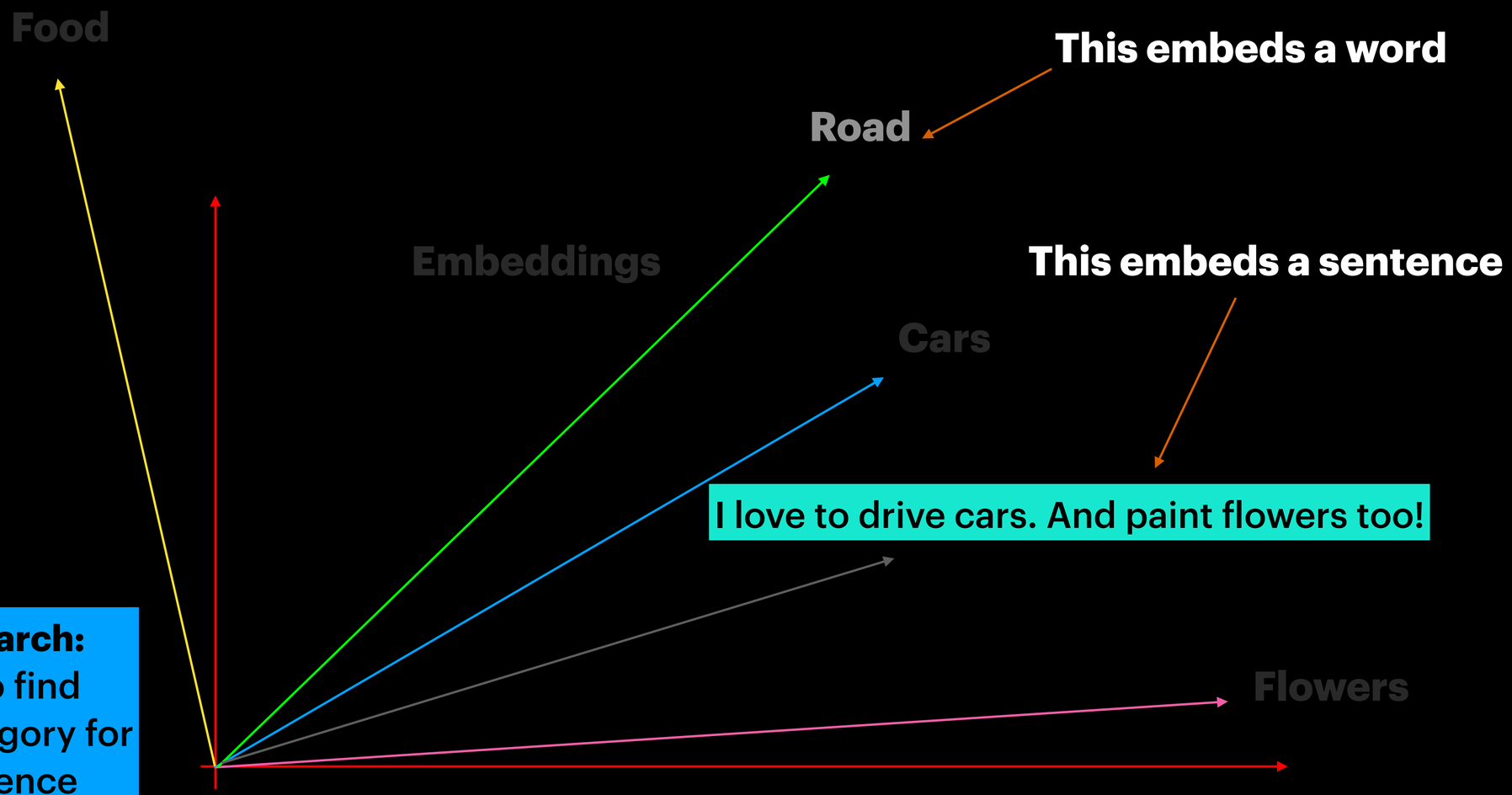


Glove Embeddings

Sentence
BERT

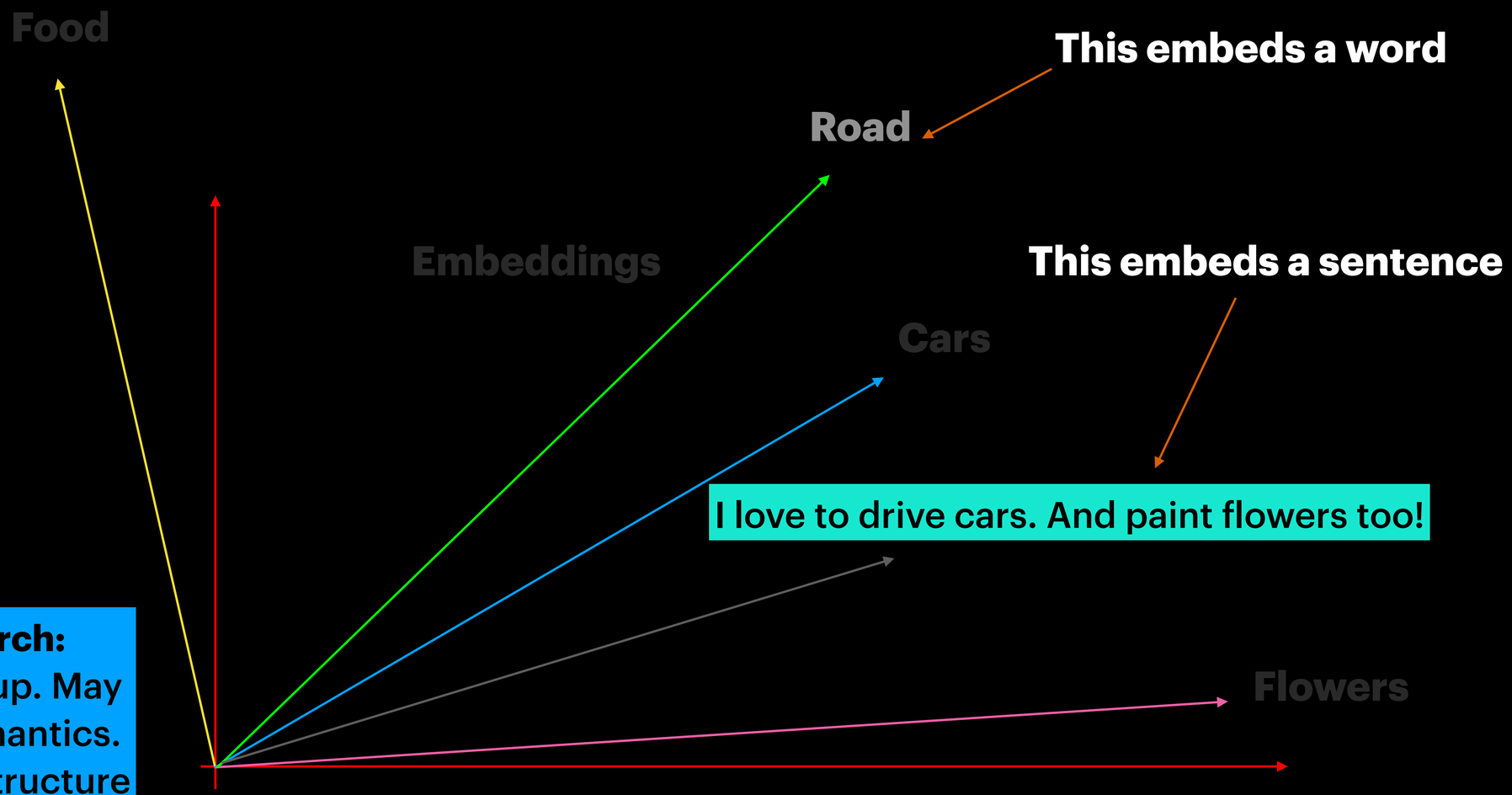
We will cover Sentence BERT when we
Get to Transformers!

Semantic Search | Vector Search



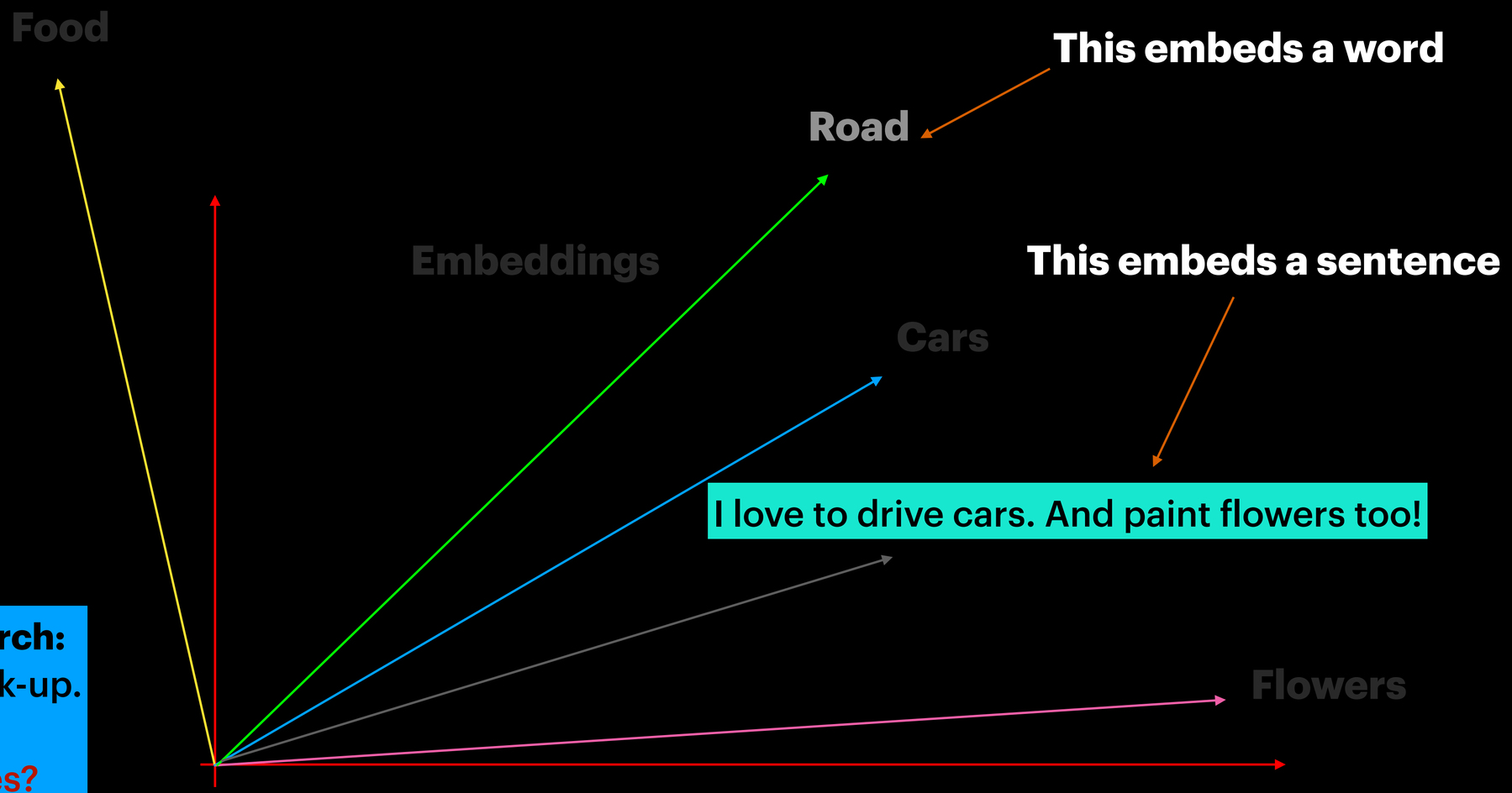
Semantic Search:
Enables us to find
the closest category for
A given sentence

Semantic Search | Vector Search



Typical Search:
Based on look-up. May Not handle semantics. Uses Trie Datastructure

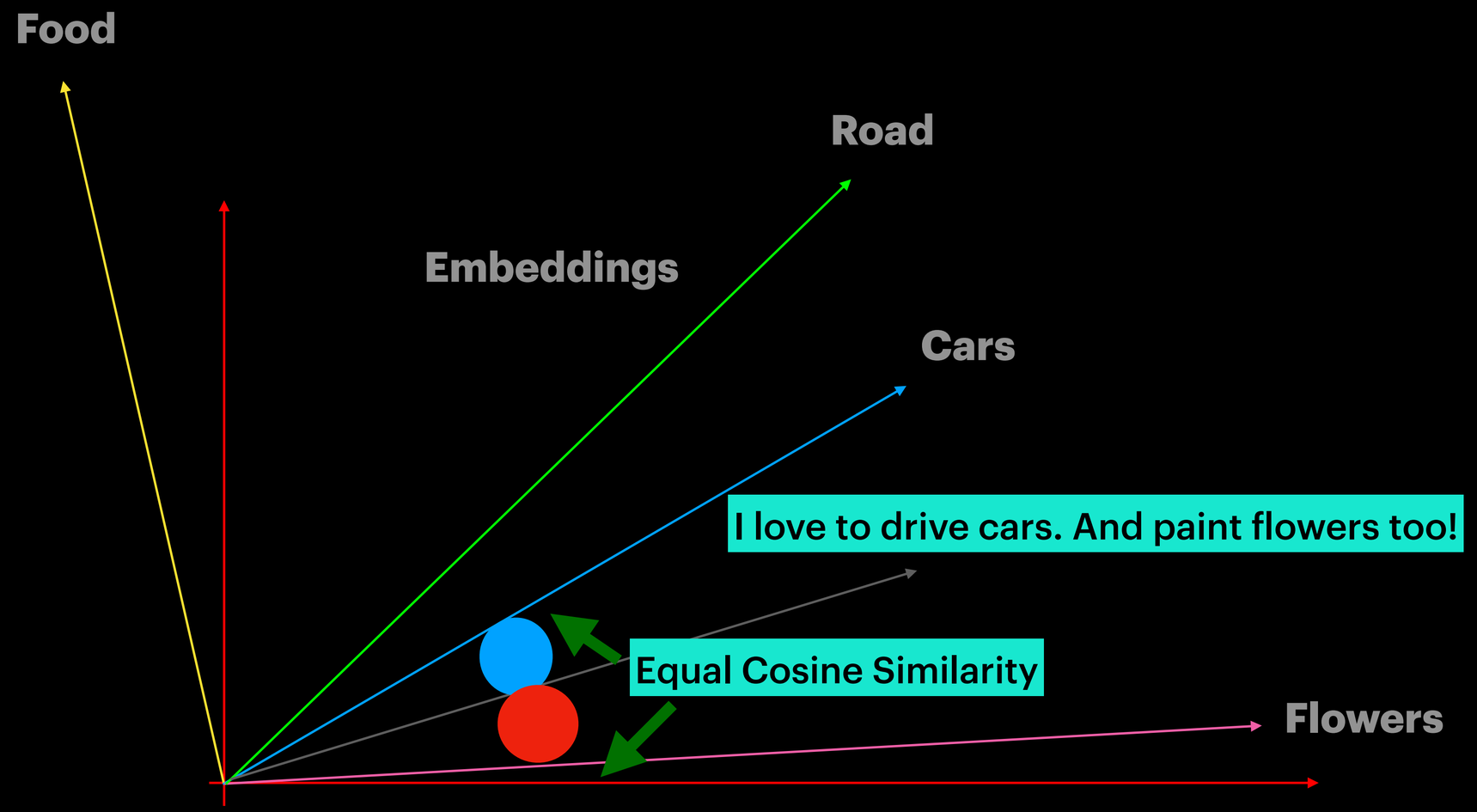
Semantic Search | Vector Search



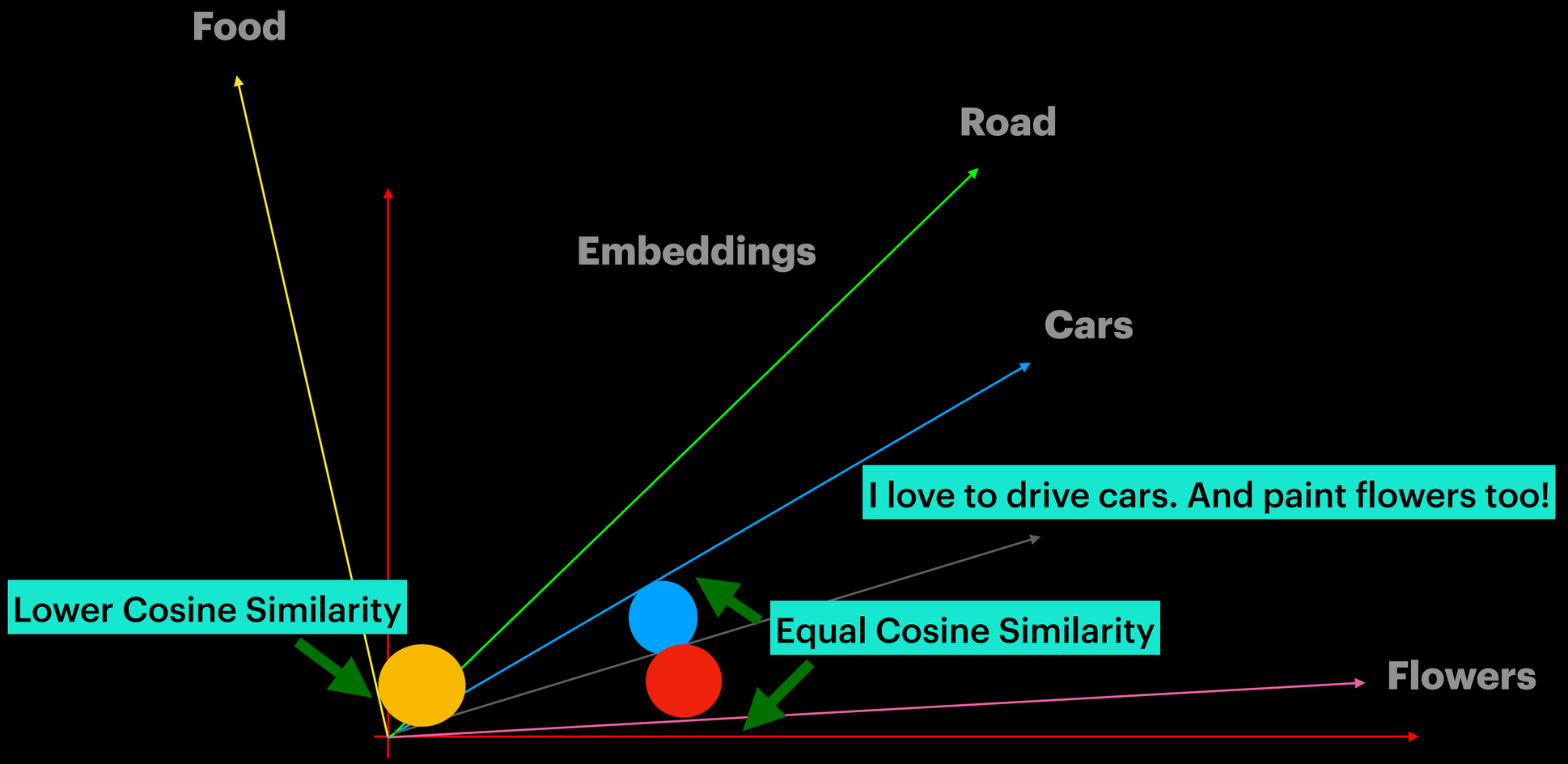
Typical Search:
Based on look-up.

Down-sides?

Semantic Search | Vector Search

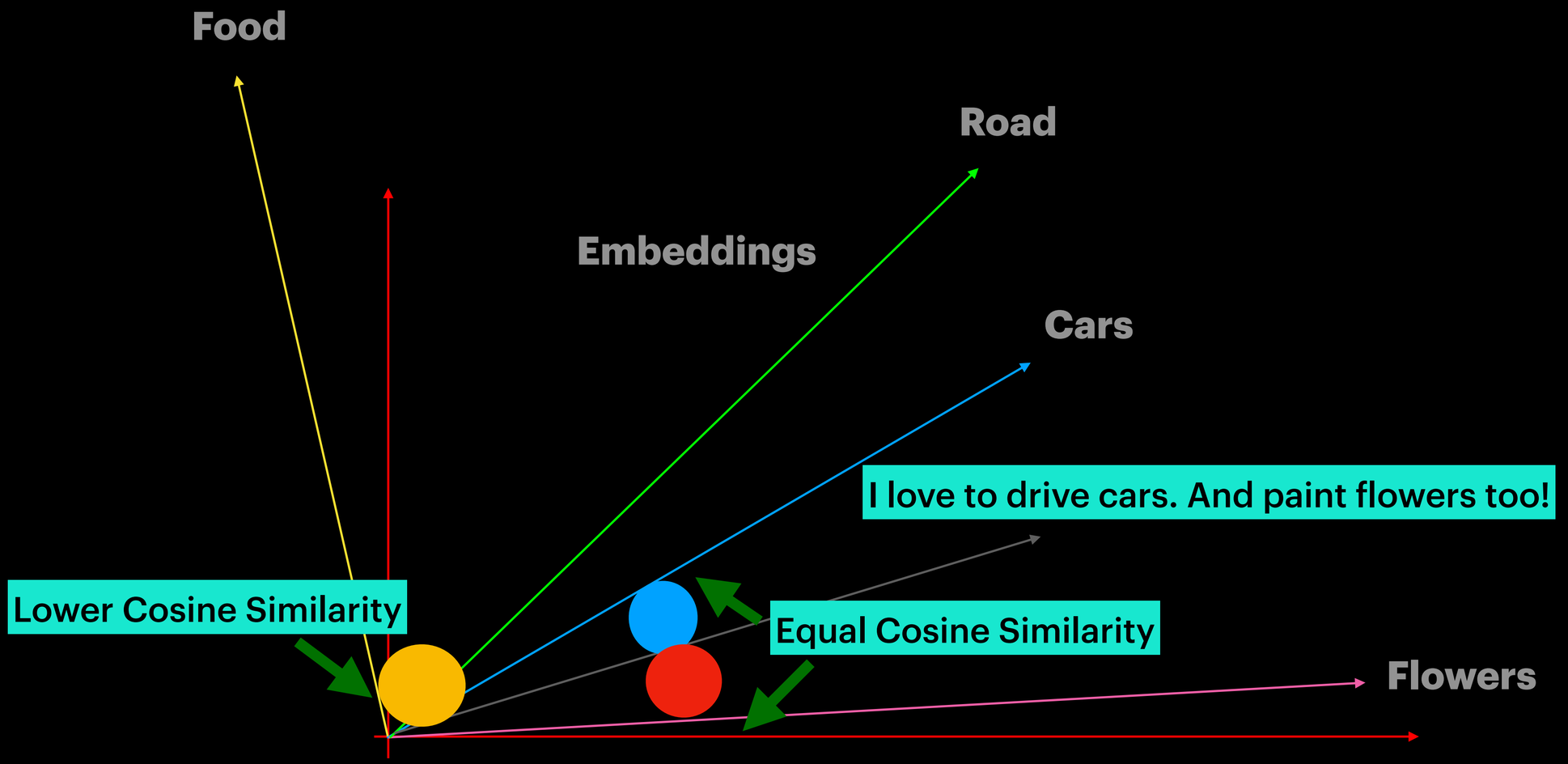


Semantic Search | Vector Search



What's the closest category
for the following sentence? "I love to drive cars. And paint flowers too!"

Semantic Search | Vector Search



What's the closest category
for the following sentence? "I love to drive cars. And paint flowers too!"

Highest cosine similarity based on vector search: Flowers and Cars

Vector Arithmetic!

What is King - Man + Woman?

Demo on Semantic Search