



THE MODERN ARCHITECTURE OF SEARCH

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1500+ Brands



20 Million
Customers



15 European
Countries





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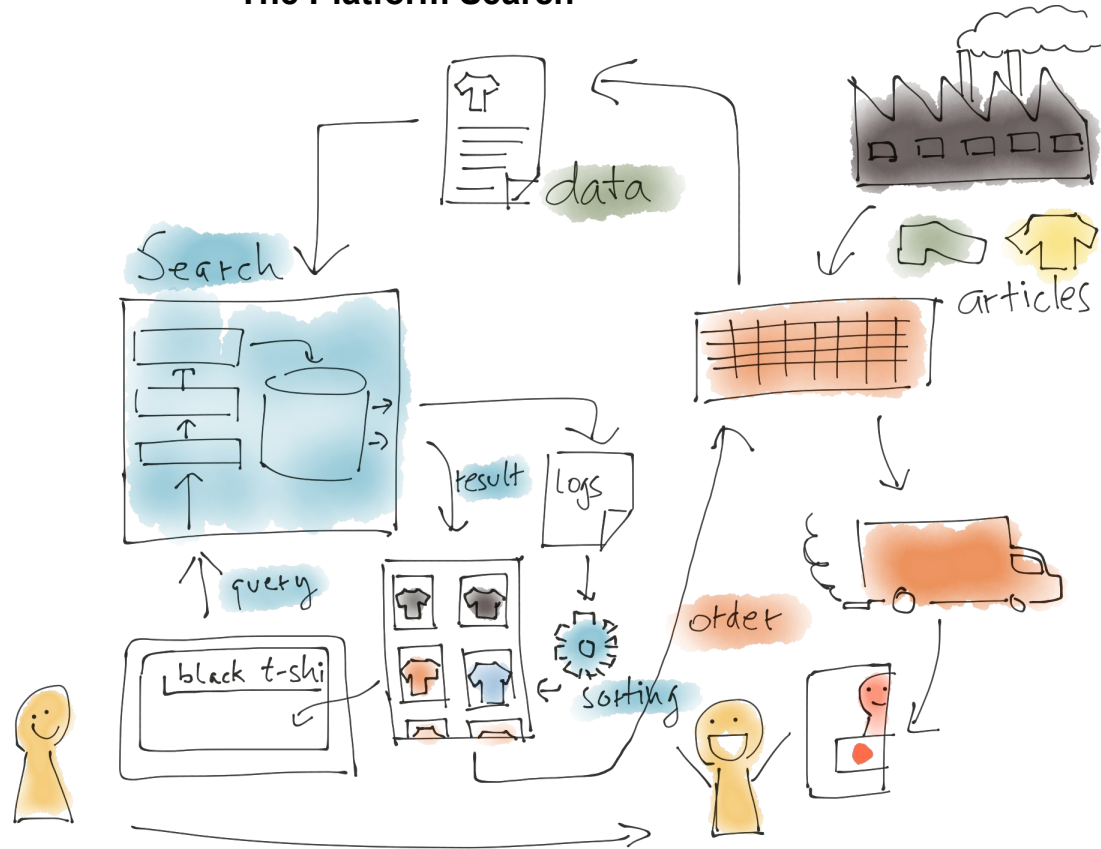
Delivery Lead

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🐦 @mikiobraun

The Platform Search

The platform search is a gateway offering search as a service for merchants and retailers. Its current focus is integration with its first and main stakeholder the fashion store.



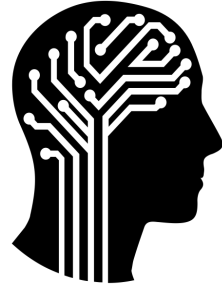


OLD SKOOL SEARCH

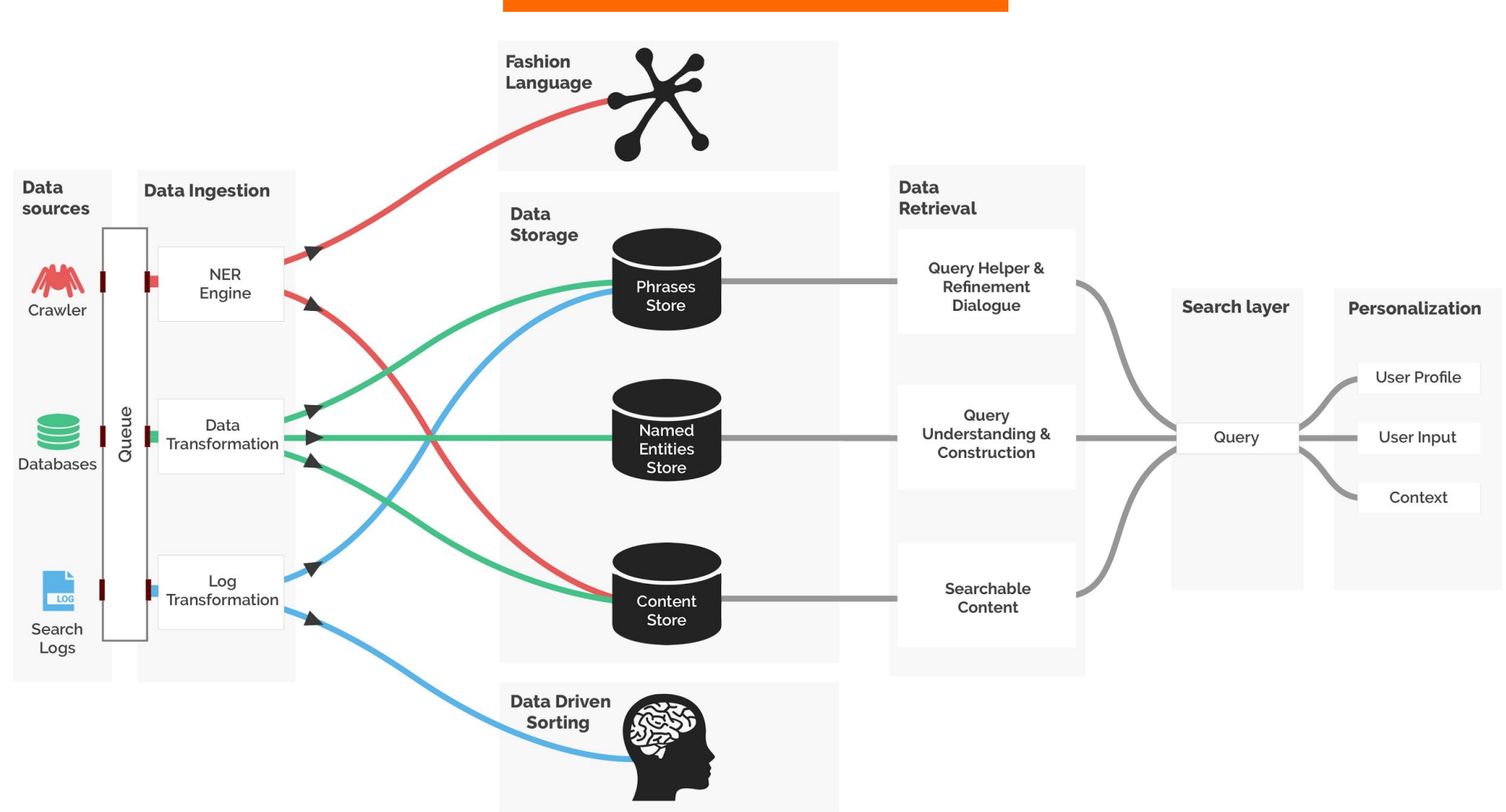
STRING MATCHING

White adidas performance sneakers 🔍



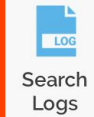


A DATA-DRIVEN WORLD



DATA INGESTION

Data sources



Crawler

Databases

Search Logs

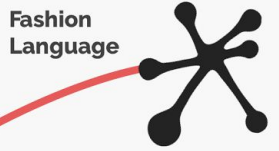
Queue

Data Ingestion

NER Engine

Data Transformation

Log Transformation



Data Storage



Data Retrieval

Query Helper & Refinement Dialogue

Query Understanding & Construction

Searchable Content

Search layer

Query

Personalization

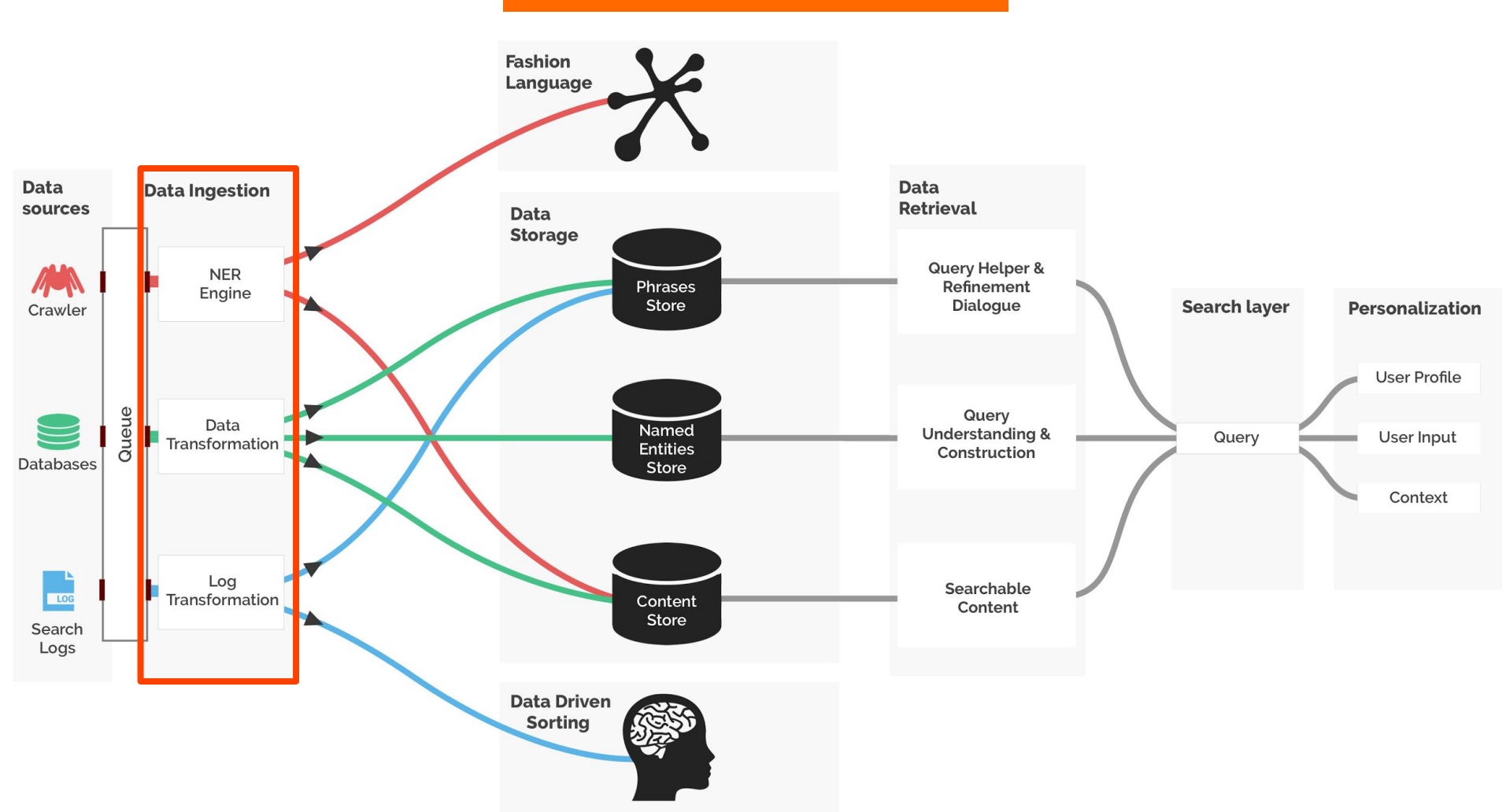
User Profile

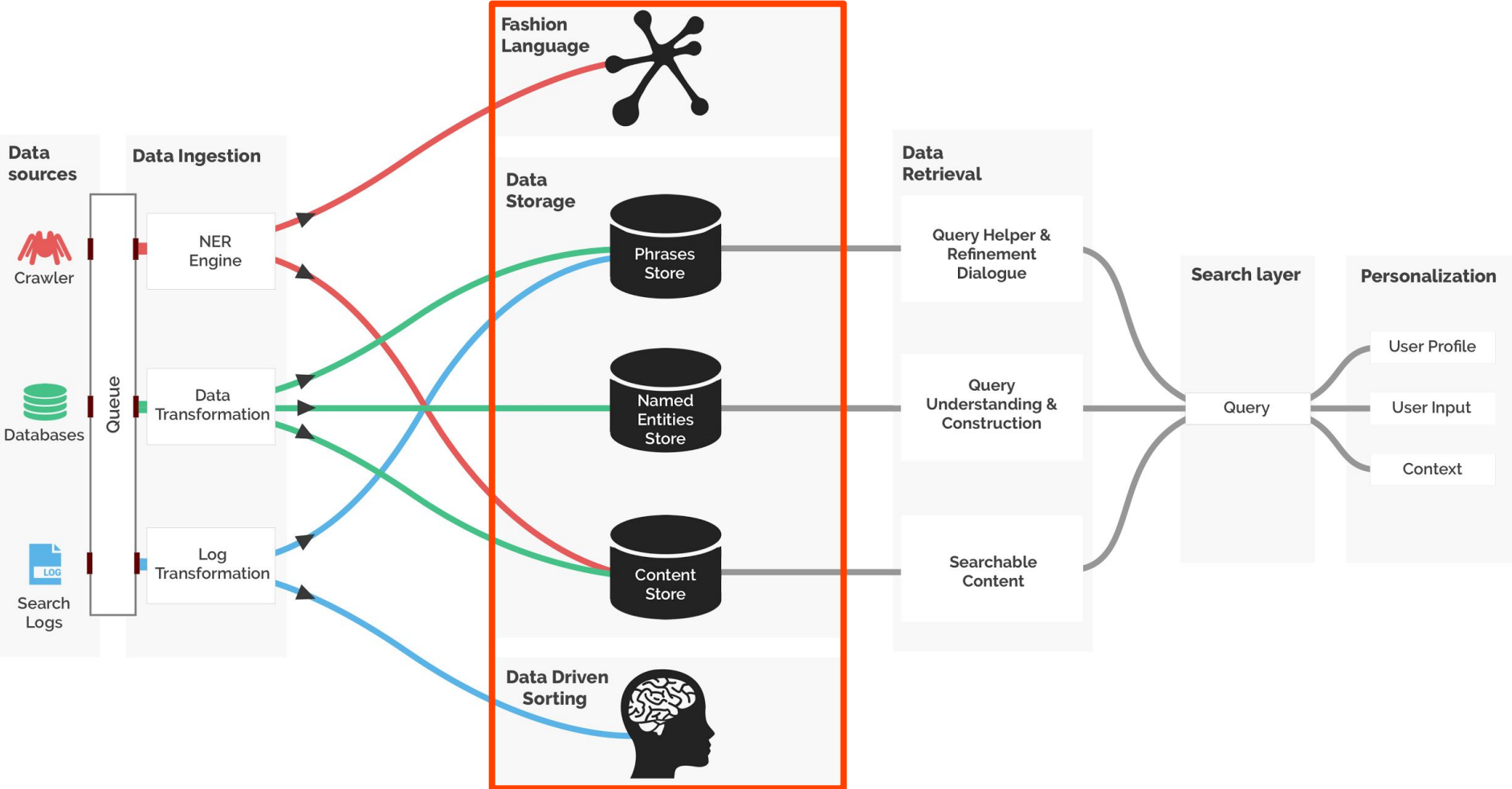
User Input

Context

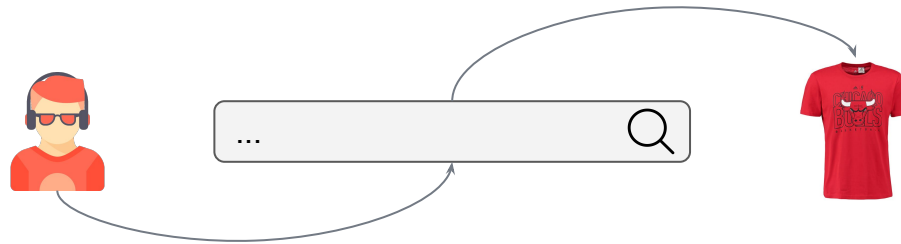


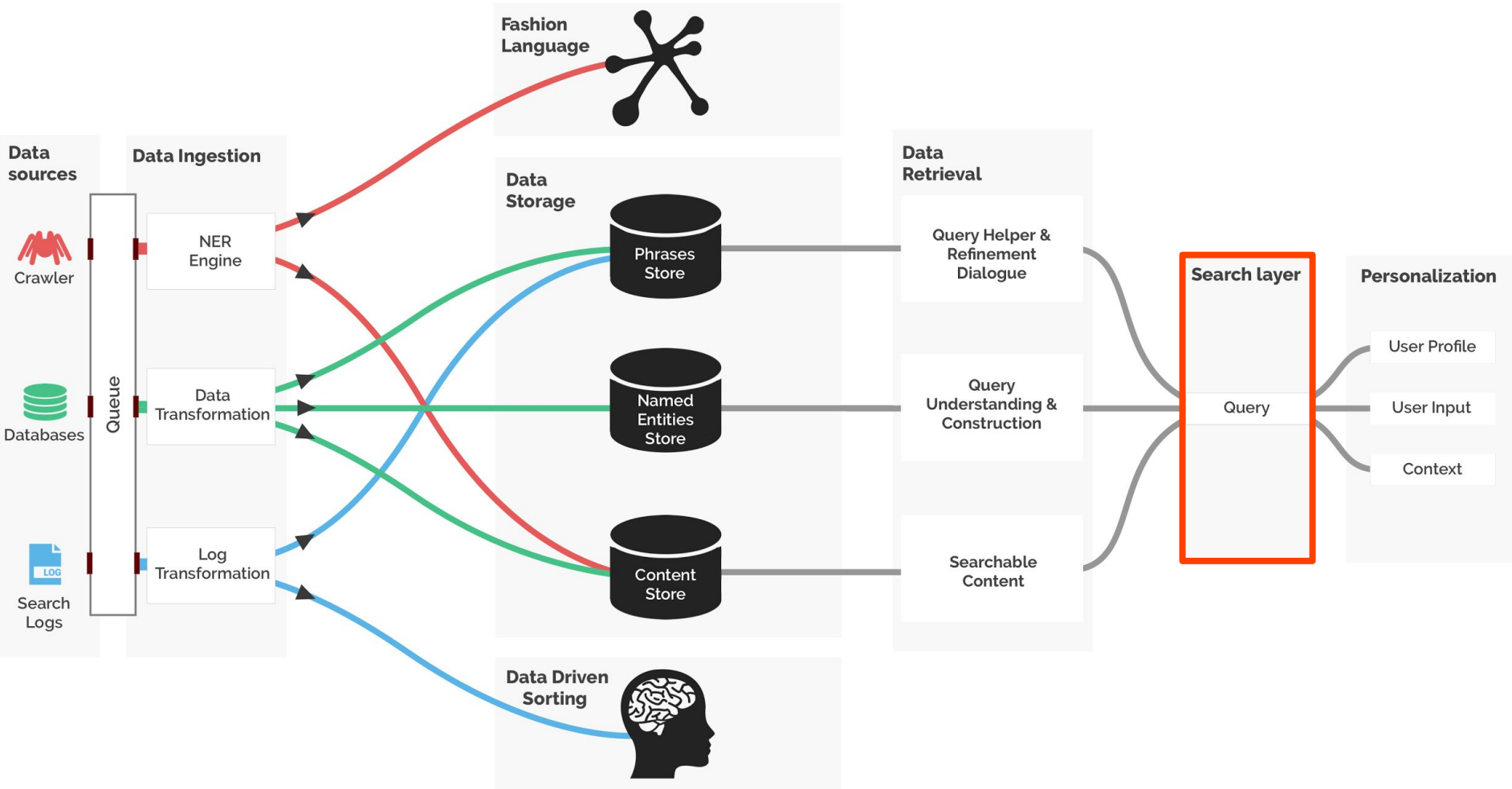
Data Driven Sorting



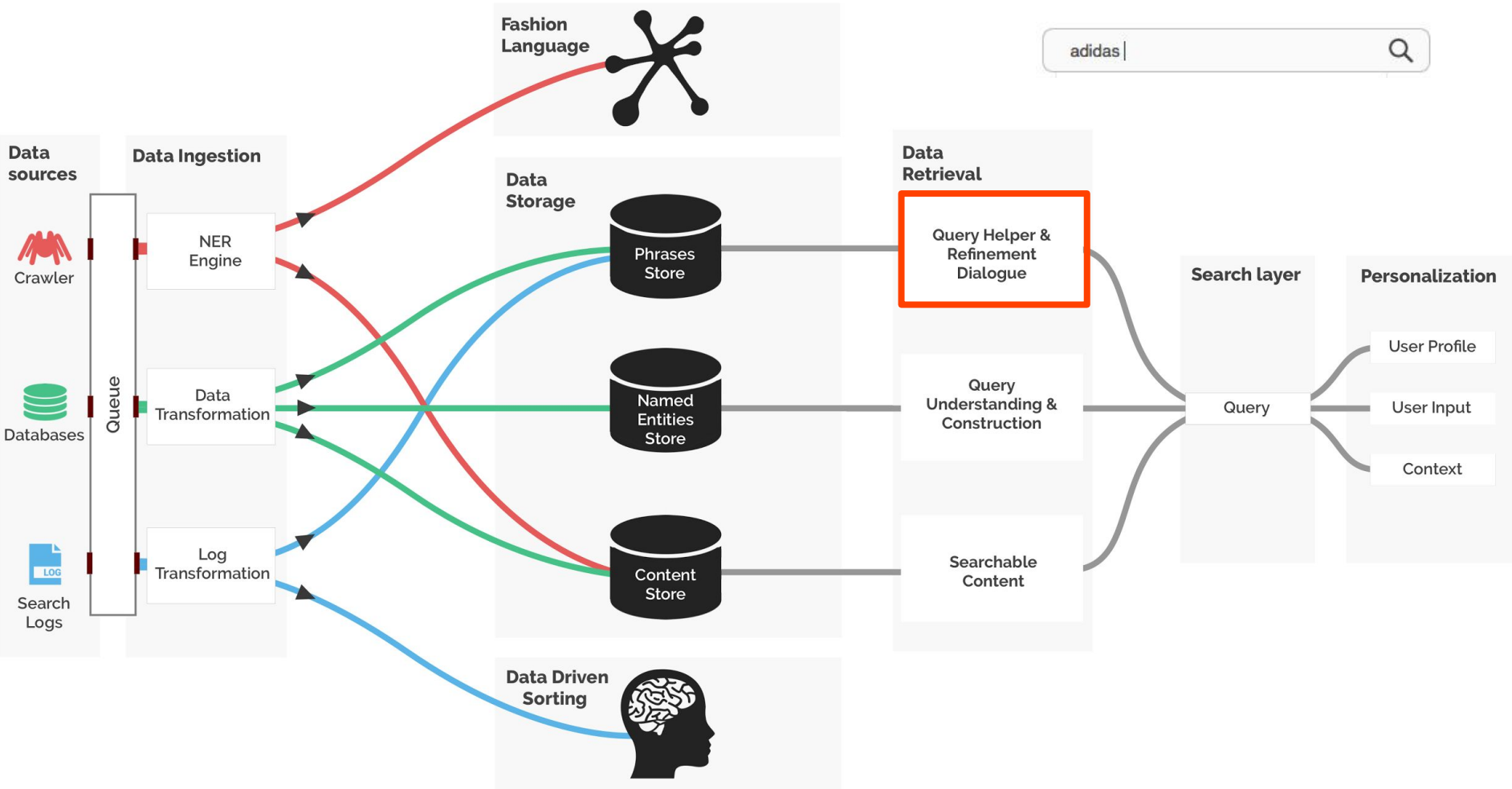


THE STORY OF A QUERY

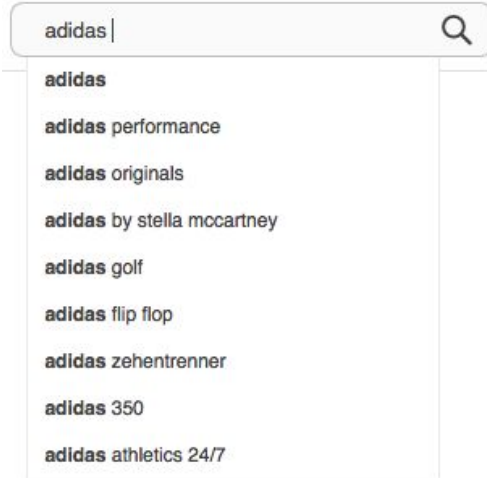




adidas | 

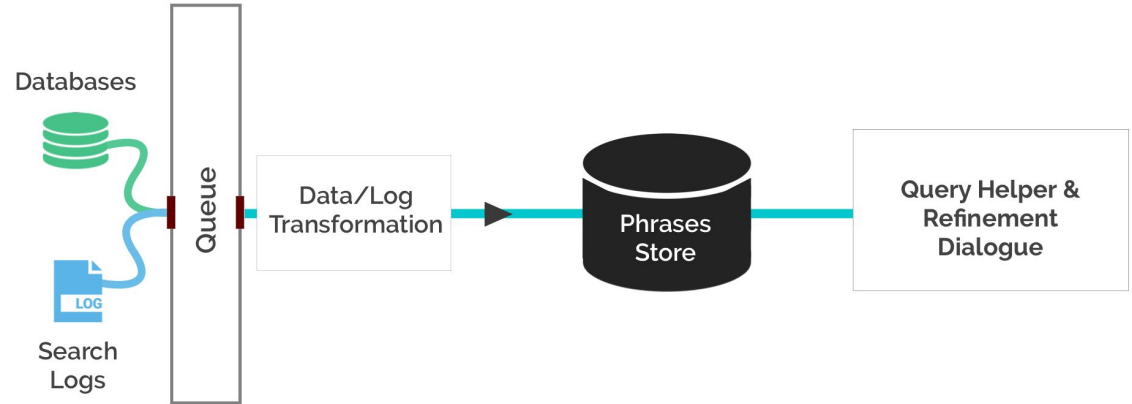
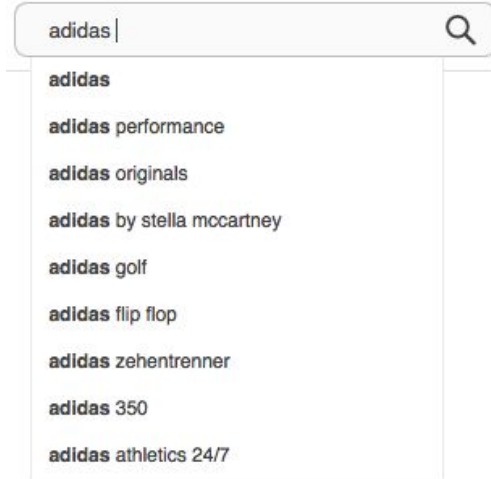


AUTO-COMPLETION



- Correct spell-error
- Predict the next word
- Understand synonym/acronym ("Michael Kors" \leftrightarrow "MK")
- Disambiguate ("jack" \rightarrow "Jack & Jones" or "Jacke")
- Be personalized (sensitive to the context)
- Learn from user-feedback
- Have diverse results

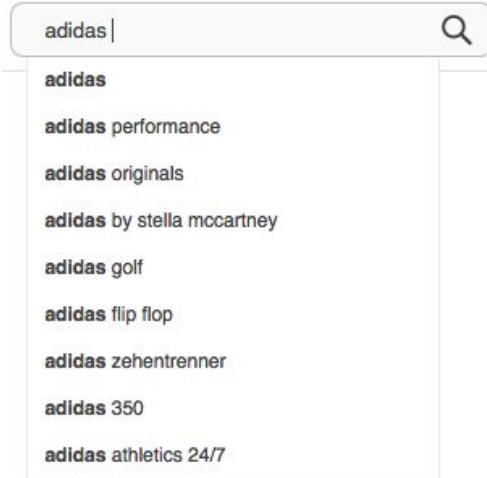
GENERATING PHRASES



Brand + Color + Material + Category + Name + =

adidas_schwarze_schuhe

RANKING PHRASES



Score:

- Hit position
- Term frequency
- Entity preference (Entity bias)
- Brevity (Penalizing long queries that match multiple entities)
- User click (click count and recency bias)

AUTO-SUGGEST

lange kleide



Autosuggest:

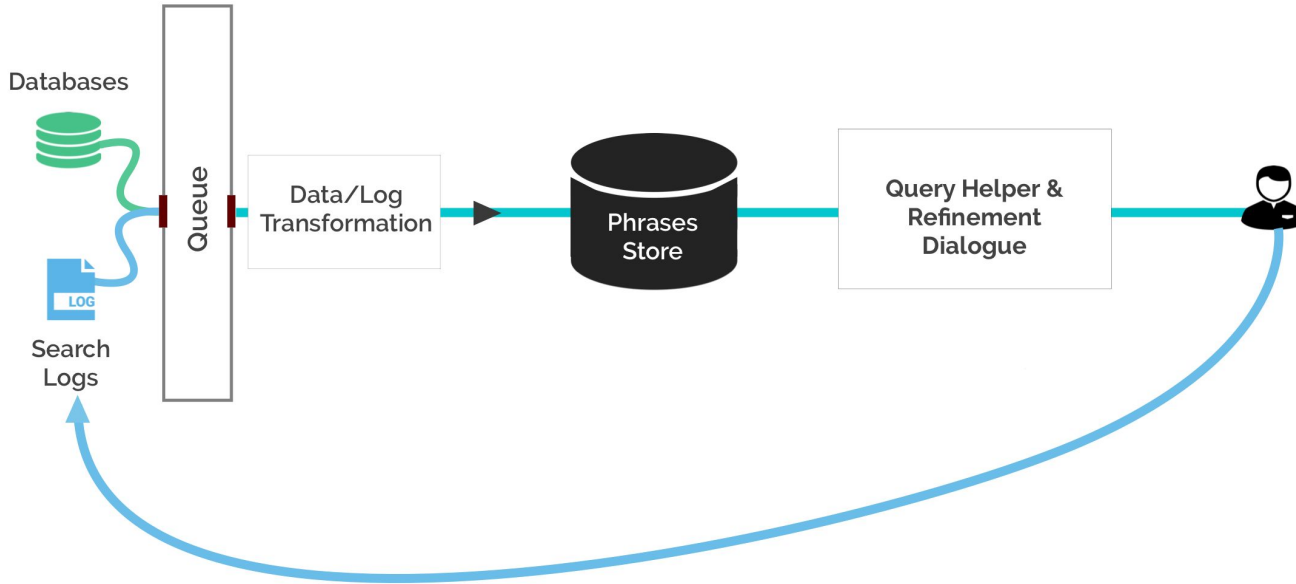
- [Did you mean ? maxikleid](#)

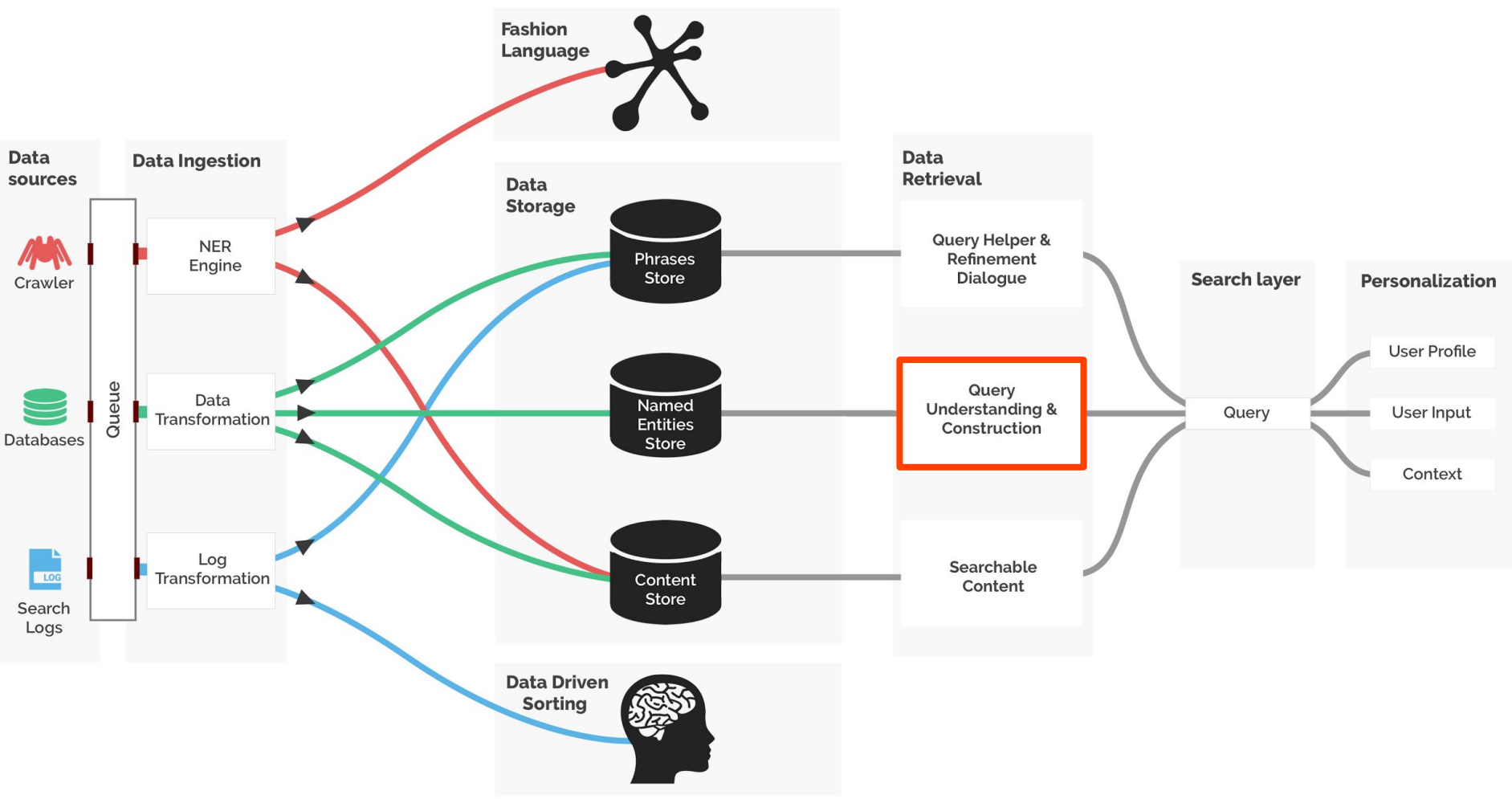
semantic space instead of text matching



maxikleid

PERSONALISATION





Amani Lederjacken 🔍

“Armani Lederjacken”

Spellcheck

“No”

Redirect

leder: CATEGORY 0.8, MATERIAL 0.4

NER



Interpretation

Armani: BRAND
Leder: MATERIAL
Jacke: CATEGORY

Armani: BRAND
Lederjacke: CATEGORY

Armani: NAME
Leder: MATERIAL
Jacke: CATEGORY

Armani: NAME
Leder: FALLBACK
Jacke: FALLBACK

Enrich

Enrich

Enrich

Enrich

Armani, Armani Jeans:
BRAND
Leder, Kunstleder:
MATERIAL
Jacke, Wolijacken:
CATEGORY

Armani: BRAND
Lederjacke,
Bikerjacke,
Lammlederjacke,
Kunstlederjacke:
CATEGORY

Armani: NAME
Leder, Kunstleder:
MATERIAL
Jacke, Wolijacken:
CATEGORY

Armani: NAME
Leder: FALLBACK
Jacke: FALLBACK

0.9

Query

0.1

Query

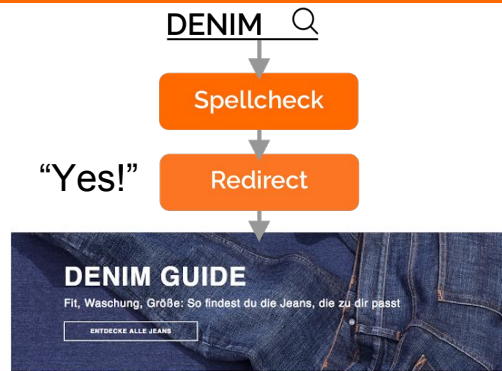
Query

Query

Scoring

Query

GUIDED SEARCH

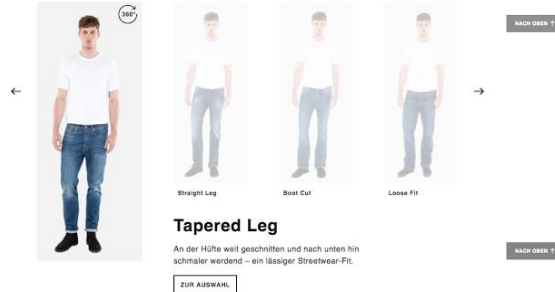


Kristof Martin, ZALON Stylist

Die passende Jeans zu finden, ist manchmal so, wie die Nadel im Heuhaufen aufzuspielen. Aber keine Panik! Ich helfe dir, dein neues Lieblingsmodell in der richtigen Größe zu bekommen. Lass dich bei ZALON von mir beraten!
[Entdecke ZALON](#)

JEANS FITS

Passformen für jeden Geschmack



Tapered Leg

An der Hüfte weit geschnitten und nach unten hin schmaler werdend – ein lässiger Streetwear-Fit.

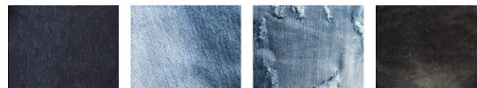
[ZUR AUSWAHL](#)

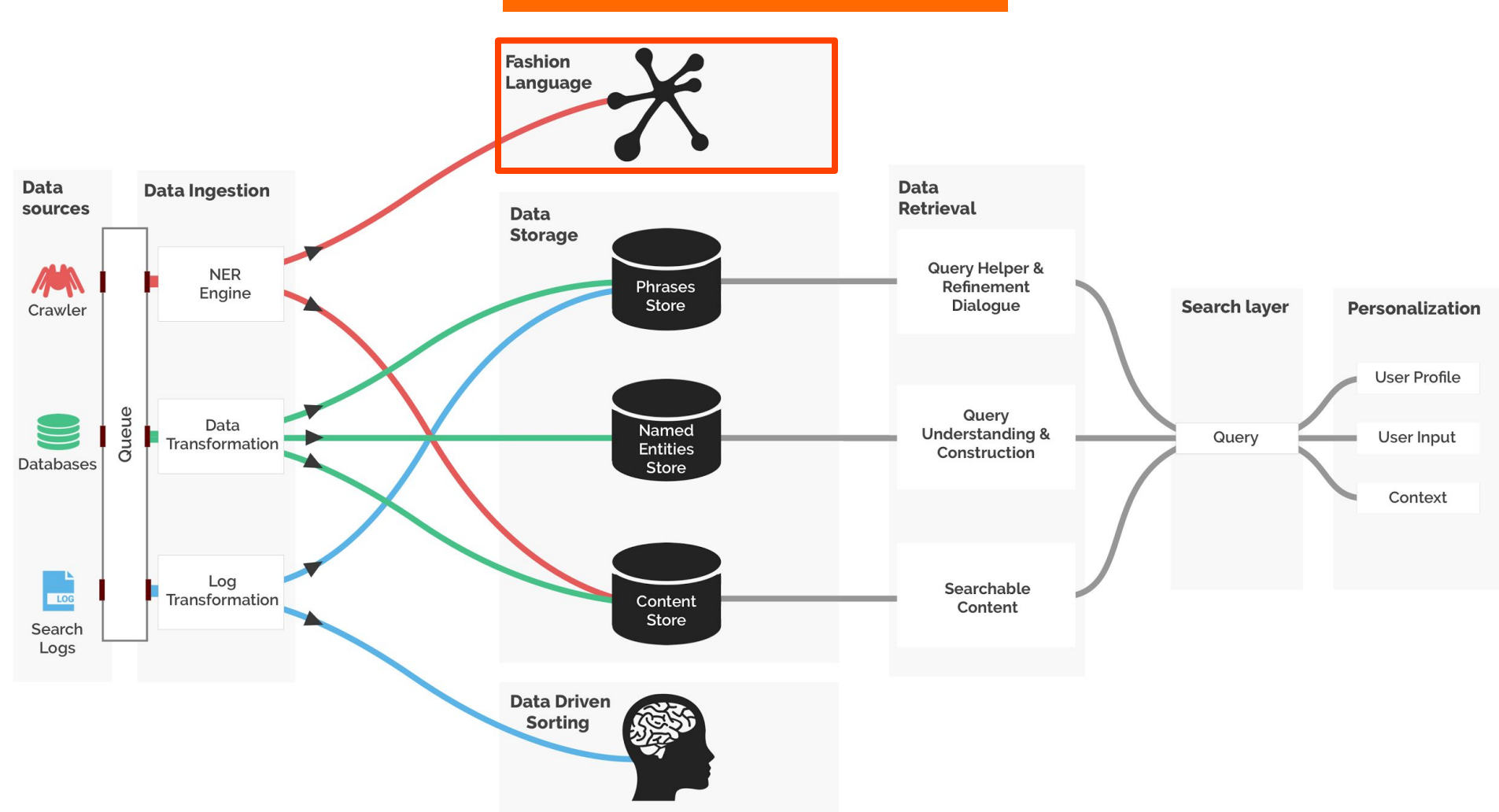
NACH OBEN 1

NACH OBEN 7

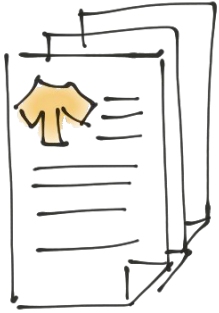
WASCHUNGEN

Finde deinen Favoriten





FASHION KNOWLEDGE BEYOND ARTICLE DATA



brands: ...

categories: ...

colors: ...

Article data already
produces a ton of
data.

But what about...?

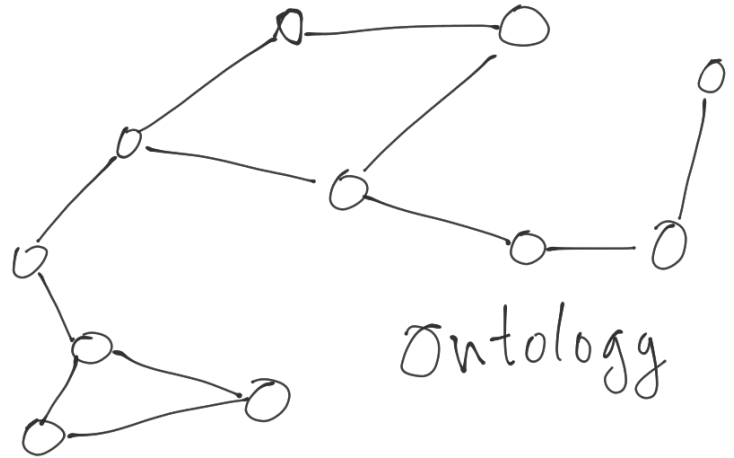
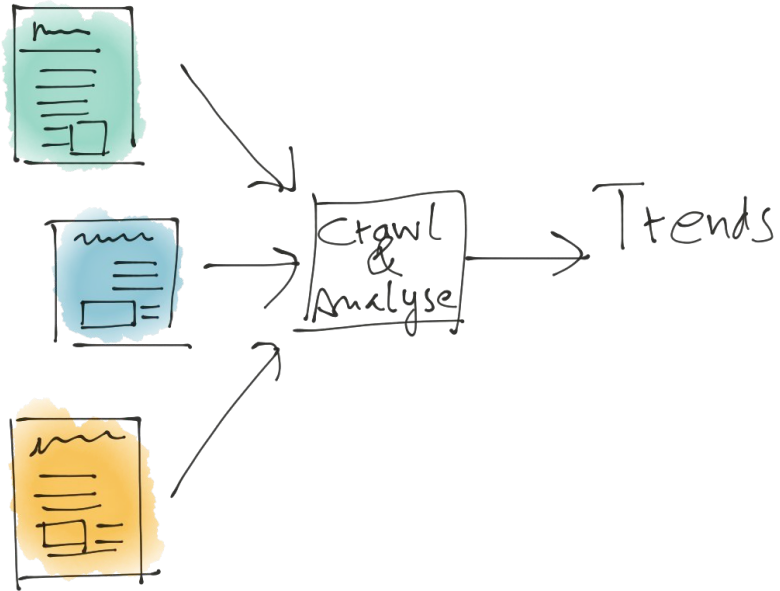
New trends

"boyfriend jeans"
"Hobo" "80s"

Occasions

"Wedding"
"Party"

FASHION KNOWLEDGE BEYOND ARTICLE DATA

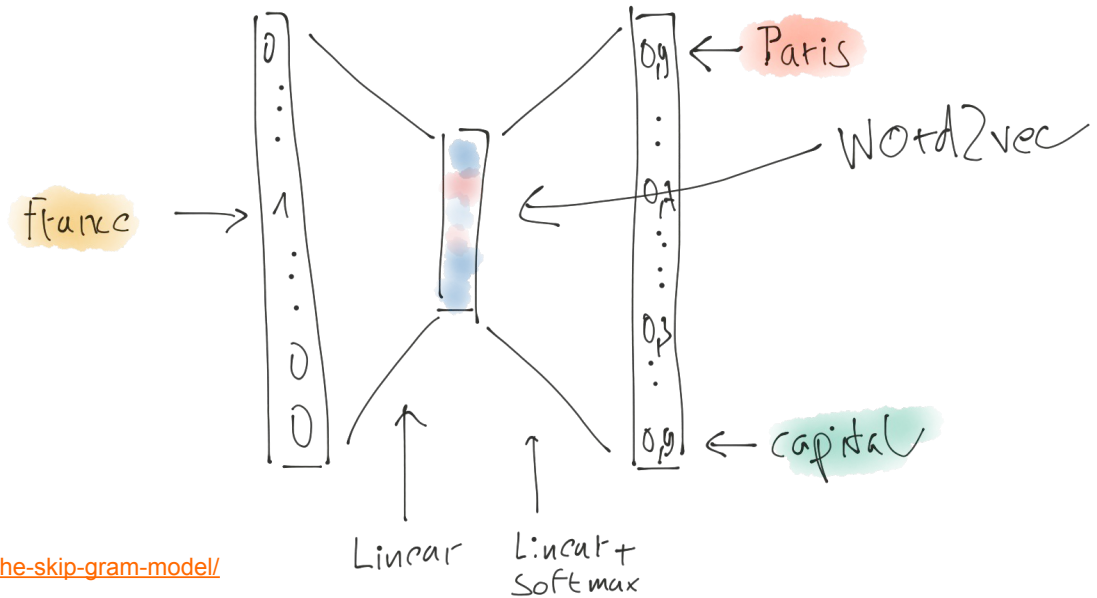


DATA DRIVEN SYNONYMS - WORD2VEC

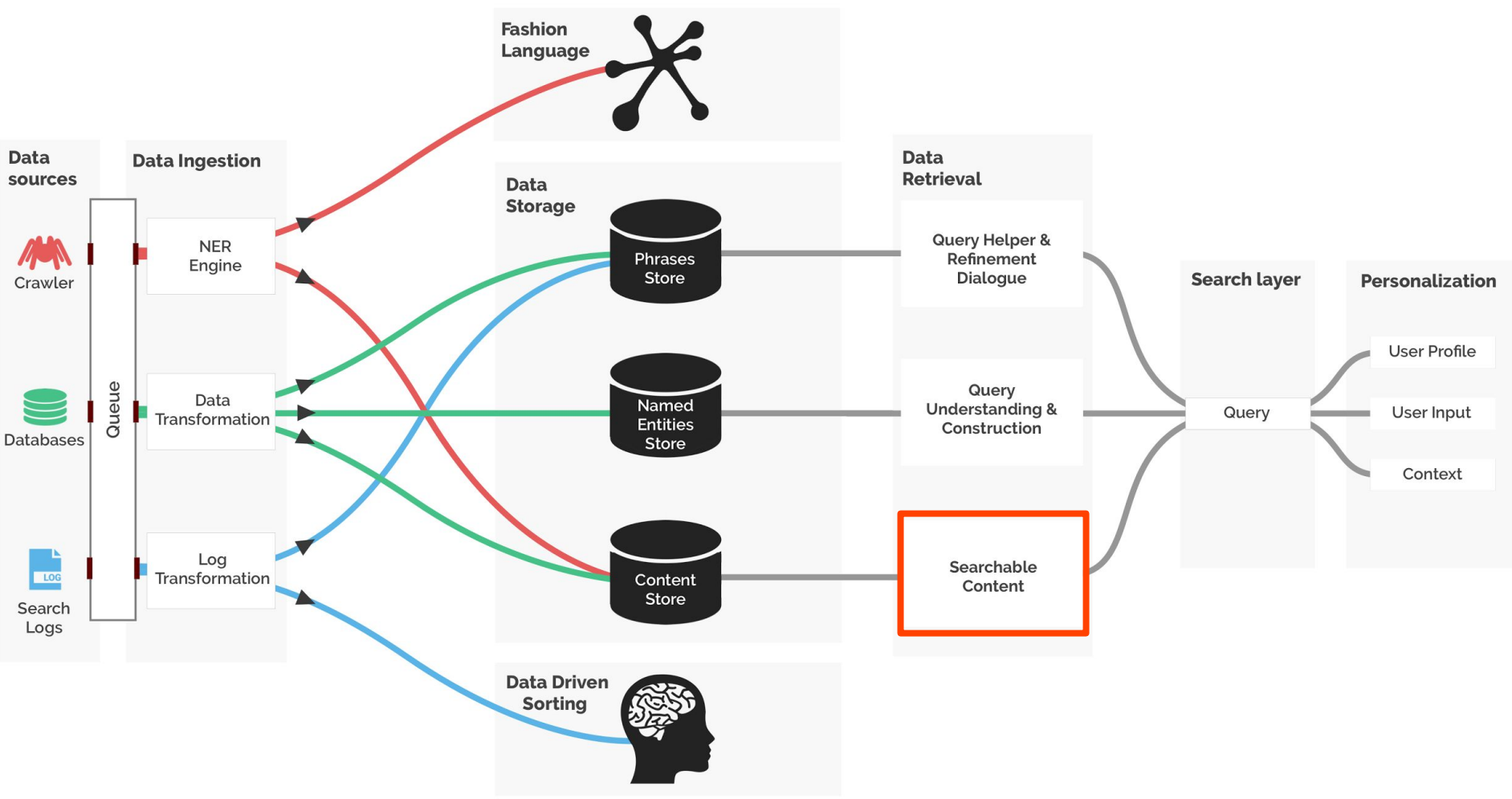
France with its capital Paris
London, the capital of England

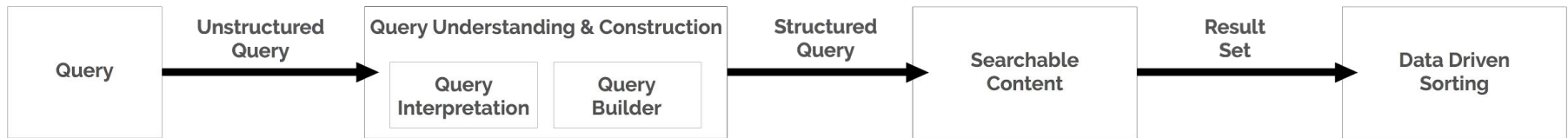
?

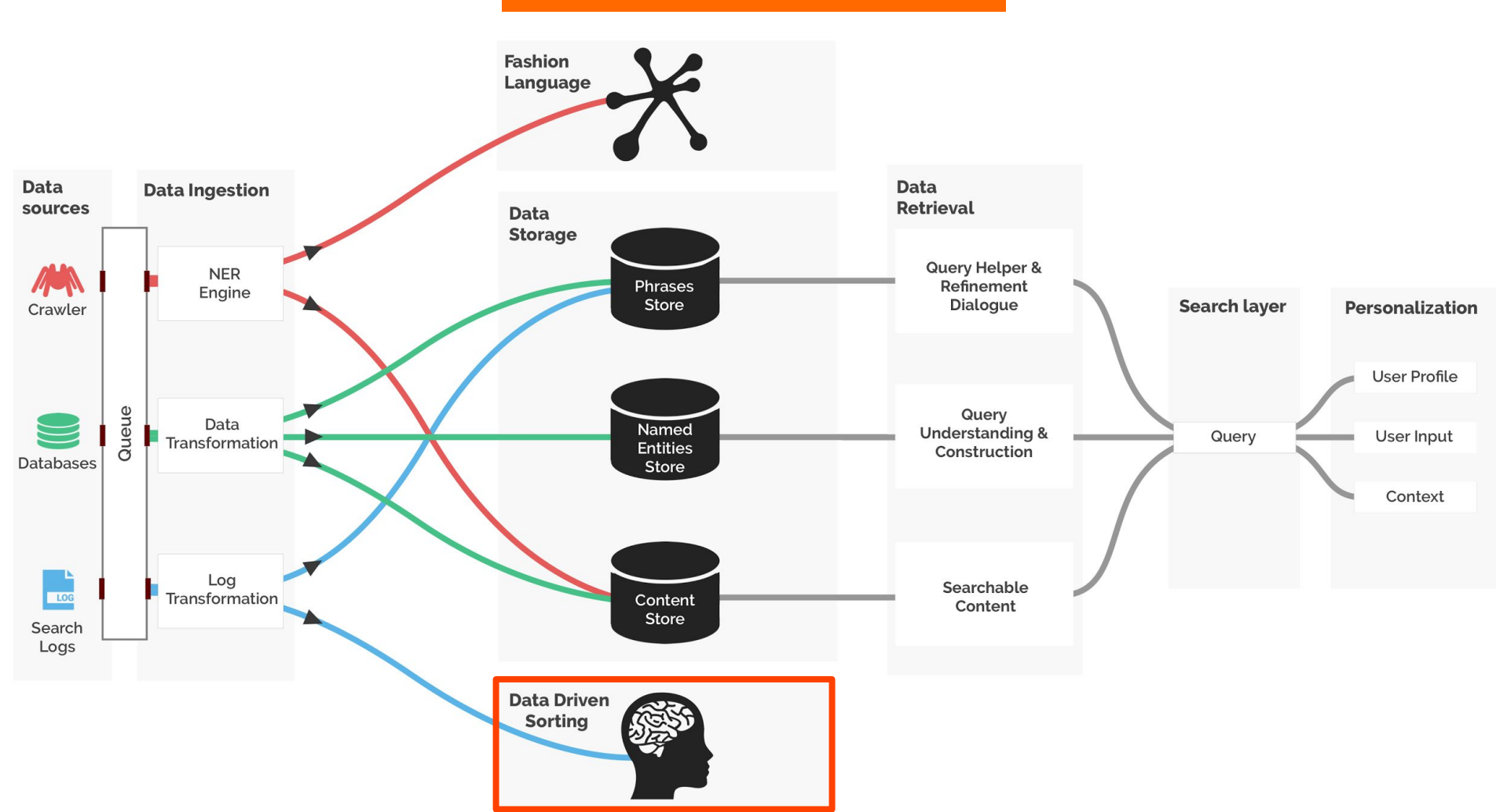
?



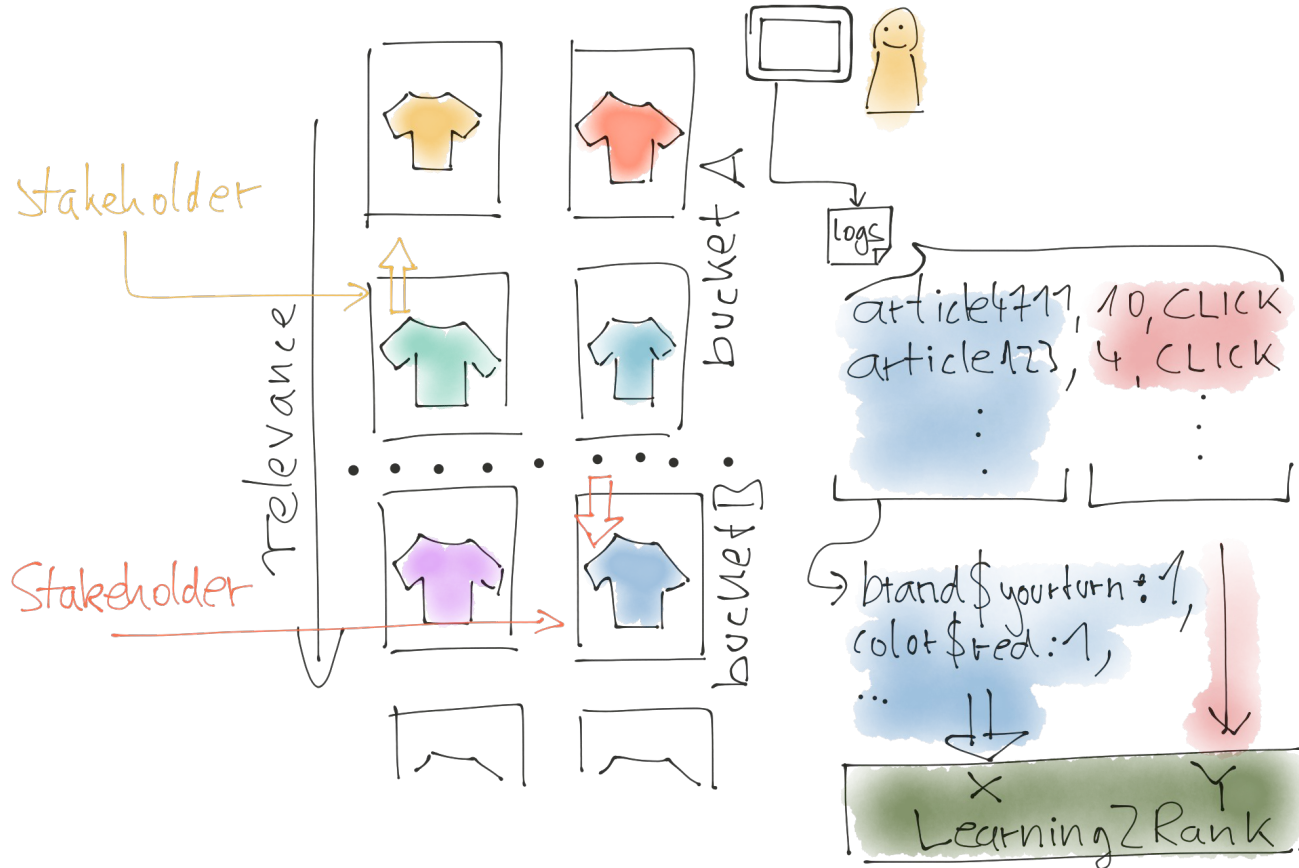
Excellent post: <http://mccormickml.com/2016/04/19/word2vec-tutorial-the-skip-gram-model/>







DATA DRIVEN SORTING



DATA DRIVEN SORTING: LEARNING TO RANK

Antonio Freno et al. "One-Pass Ranking Models for Low-Latency Product Recommendations", KDD 2015

Dataset: (l, r, \mathcal{X})

l_i = relevance for user

r_i = rank $1 \dots n$

\mathcal{X}_i = feature for item i

$$\varphi_w = X^T w$$

$$\min_w \left\{ \frac{1}{m} \sum_{j=1}^m \ell_{\mathcal{M}(r^{(j)}, l^{(j)})}(X^{(j)}; \varphi_w) + \lambda_1 \|w\|_1 + \frac{1}{2} \lambda_2 \|w\|_2^2 \right\}$$

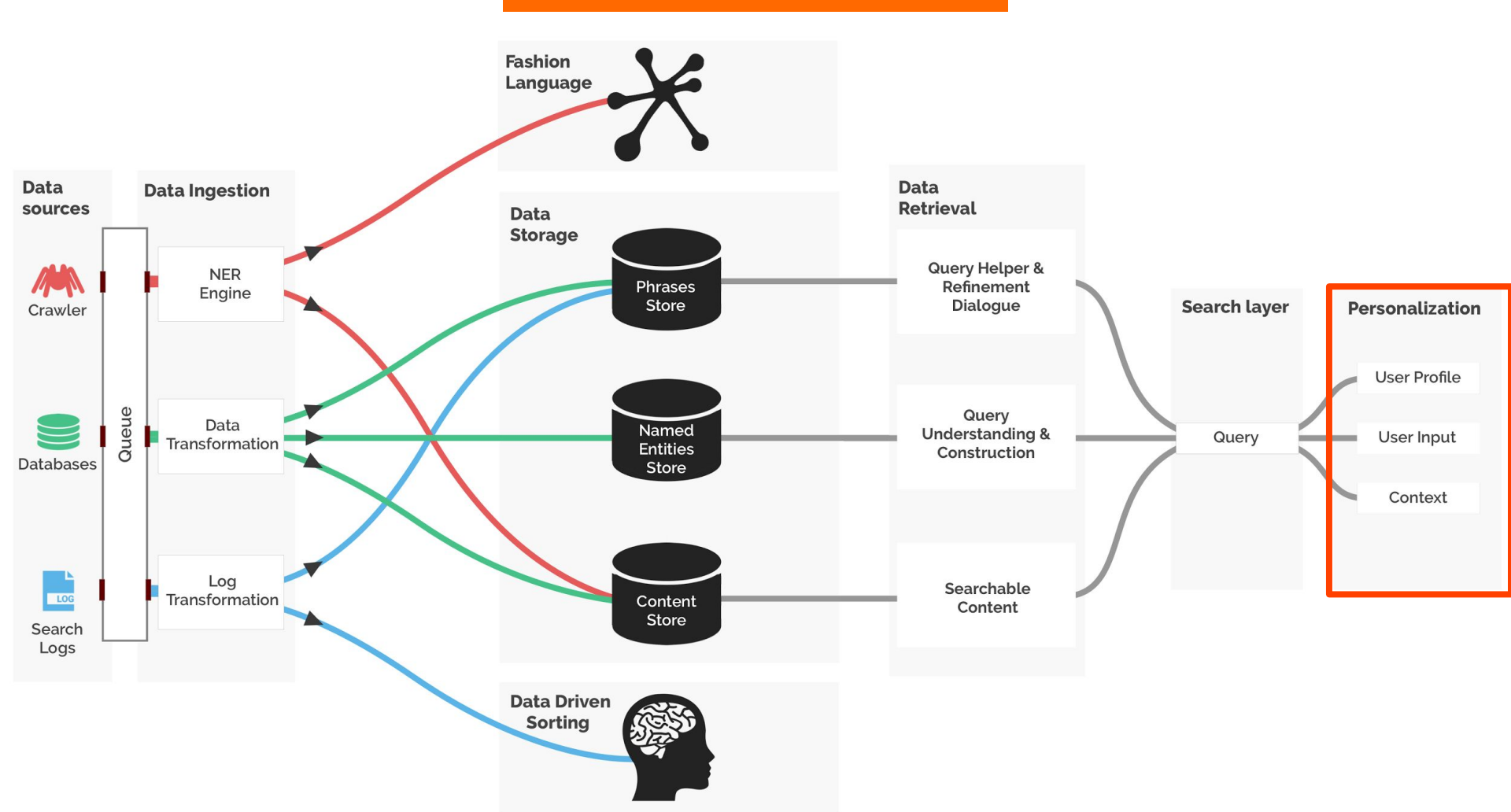
$$\ell_{\mathcal{M}(r, l)}(X; \varphi) = \sum_{r_i \leq r_j} \Delta_{\mathcal{M}(r, l)}(i, j) \mathcal{P}(\varphi(x_i), \varphi(x_j))$$

$$\Delta_{\mathcal{M}(r, l)}(i, j) = \mathcal{M}(r, l) - \mathcal{M}(r_{i \setminus j}, l)$$

tanking metric

pairwise metric (hinge)

difference by exchanging i, j



PERSONALISATION

- Goal: Find the most relevant articles for each user.
- Give the customer the feeling, search is understanding what their needs are.
- Essentially part of every step: Query Helper, Query Analysis, Personalization.
- From simple methods (history, last orders) to full blown complex models (personalized ranking models)

THE END