

Recommender Metrics Framework

Measuring the success of a Recommender System

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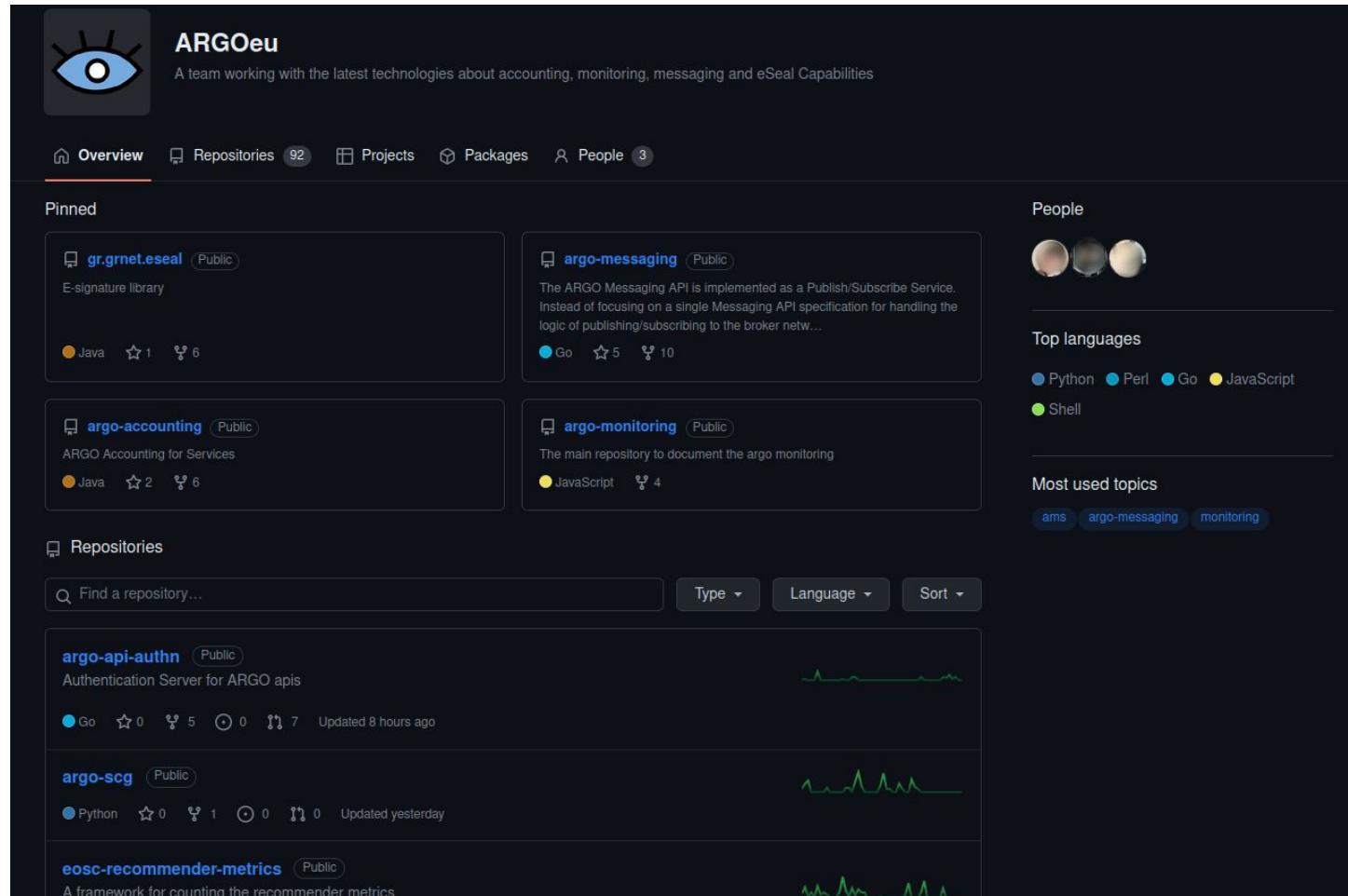
Who we are?

- GRNET S.A. – National Infrastructures for Research and Technology, is one of the largest public sector technology companies in Greece.
- It provides networking, cloud computing, HPC, data management services and e-Infrastructures and services to academic and research institutions, to educational bodies at all levels, and to all agencies of the public sector.
- <https://grnet.gr>

The screenshot shows the official website of GRNET. At the top, there is a navigation bar with links for COMPANY, SERVICES, PROJECTS, NEWS CORNER, and CONTACT, along with social media icons for Facebook, Twitter, LinkedIn, YouTube, and Instagram. A search bar is located in the top right corner. The main content area features the GRNET logo and a banner image of a smartphone resting on a book. Below this, a video player displays a thumbnail for a company video titled "Company Video GRNET | Networking Education, Research and Government". The video thumbnail includes the GRNET logo, the text "National Infrastructures for Research and Technology", and the title "GRNET | Networking Research Education and".

Who we are?

- The European Infrastructures and Projects Directorate.
- Among others, it develops software solutions in various research projects across EU and Greece.
- One of the subteams of the Directorate that lies in GitHub Organizations:
<https://github.com/ARGOeu>
- Recommender Metrics Framework (RMF) is developed by GRNET and used as an open-source solution in the EOSC-Future project.



The screenshot shows the GitHub organization page for ARGOeu. At the top, there's a header with the organization's name and a brief description: "A team working with the latest technologies about accounting, monitoring, messaging and eSeal Capabilities". Below the header, there are tabs for Overview, Repositories (92), Projects, Packages, and People (3). The Overview tab is selected. On the left, there's a sidebar with sections for People (three profile icons), Top languages (Python, Perl, Go, JavaScript, Shell), and Most used topics (ams, argo-messaging, monitoring). The main area displays four pinned repositories: gr.grnet.eseal (Public, Java, 1 star, 6 forks), argo-messaging (Public, Go, 5 stars, 10 forks), argo-accounting (Public, Java, 2 stars, 6 forks), and argo-monitoring (Public, JavaScript, 3 stars, 4 forks). Below these, there's a section for other repositories, including argo-api-authn (Public, Go, 0 stars, 5 forks, last updated 8 hours ago) and argo-scg (Public, Python, 0 stars, 1 fork, last updated yesterday). There are also two repos under eosc-recommender-metrics (Public, Go, 0 stars, 0 forks, last updated 1 day ago).

What are we going to talk about?

A Recommender System's Metrics Framework that can produce measurable results for a Recommender System's evaluation.



Where is it used?

Will explain in a few minutes

To monitor and report diagnostic metrics for the EOSC Marketplace Recommender Service.

Recommender Service

- Owners
- Developers
- Engineers

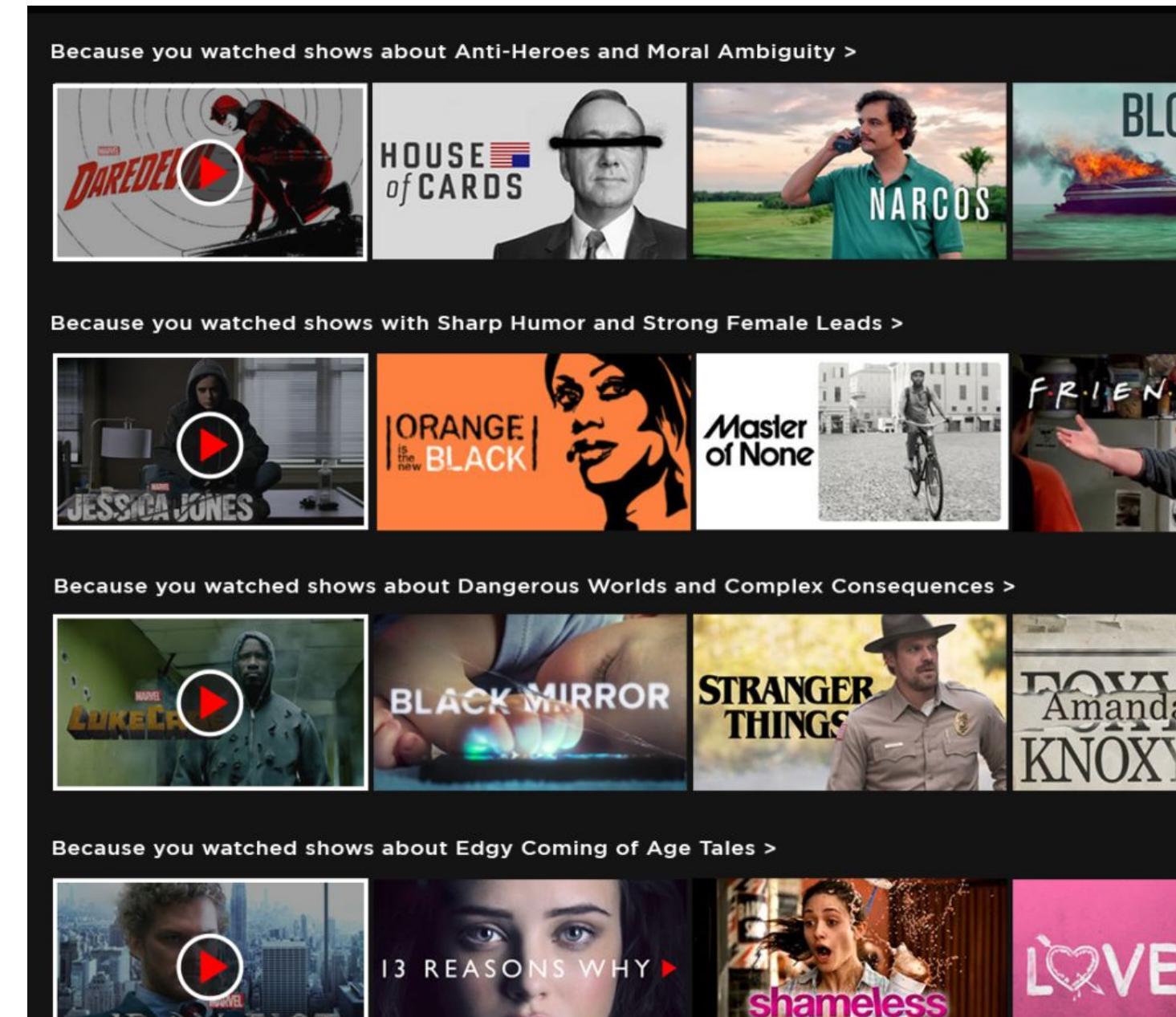


Service using the RS Marketing Team



What is a Recommender System (RS)?

- Offers personalized suggestions to users.
- Recommendations are based on user preferences, behaviors, patterns.
- Recommendations can include products, content, services, or connections.
- Aims to enhance user experience.
- Applications include e-commerce, content streaming, and social media.
- Addresses the information overload problem in the digital age.



eoscfuture.eu



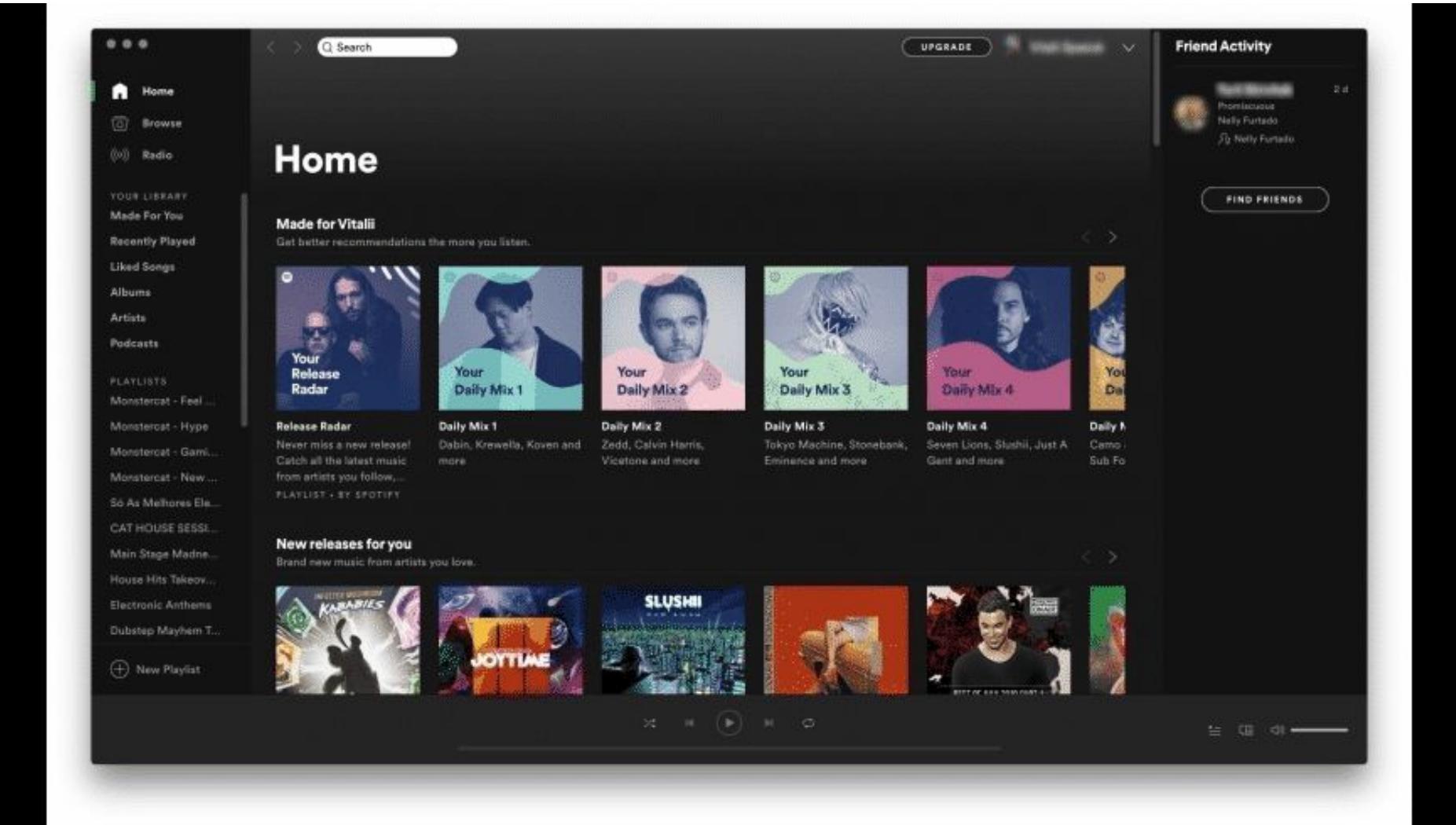
@EOSCFuture



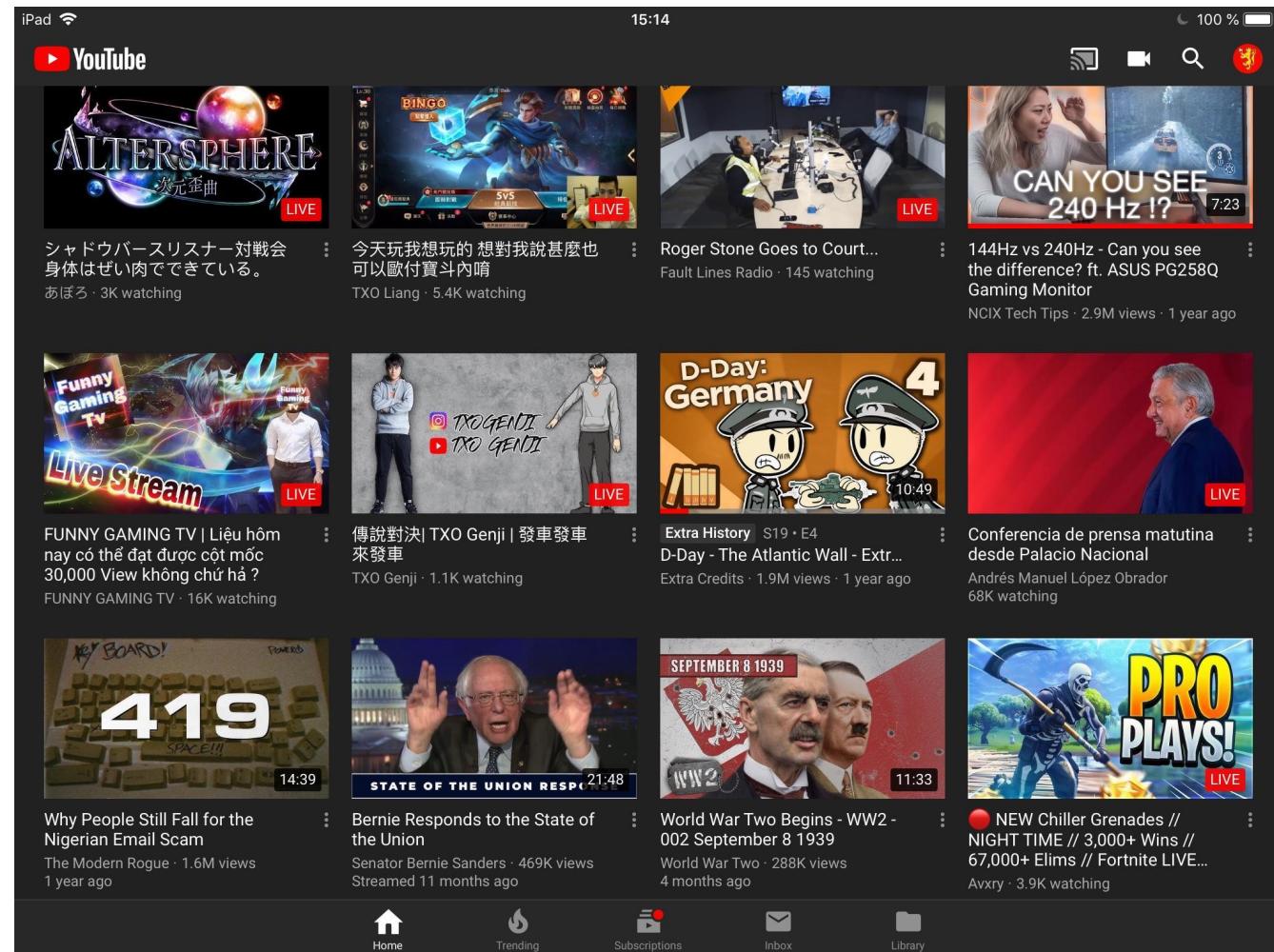
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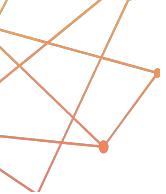


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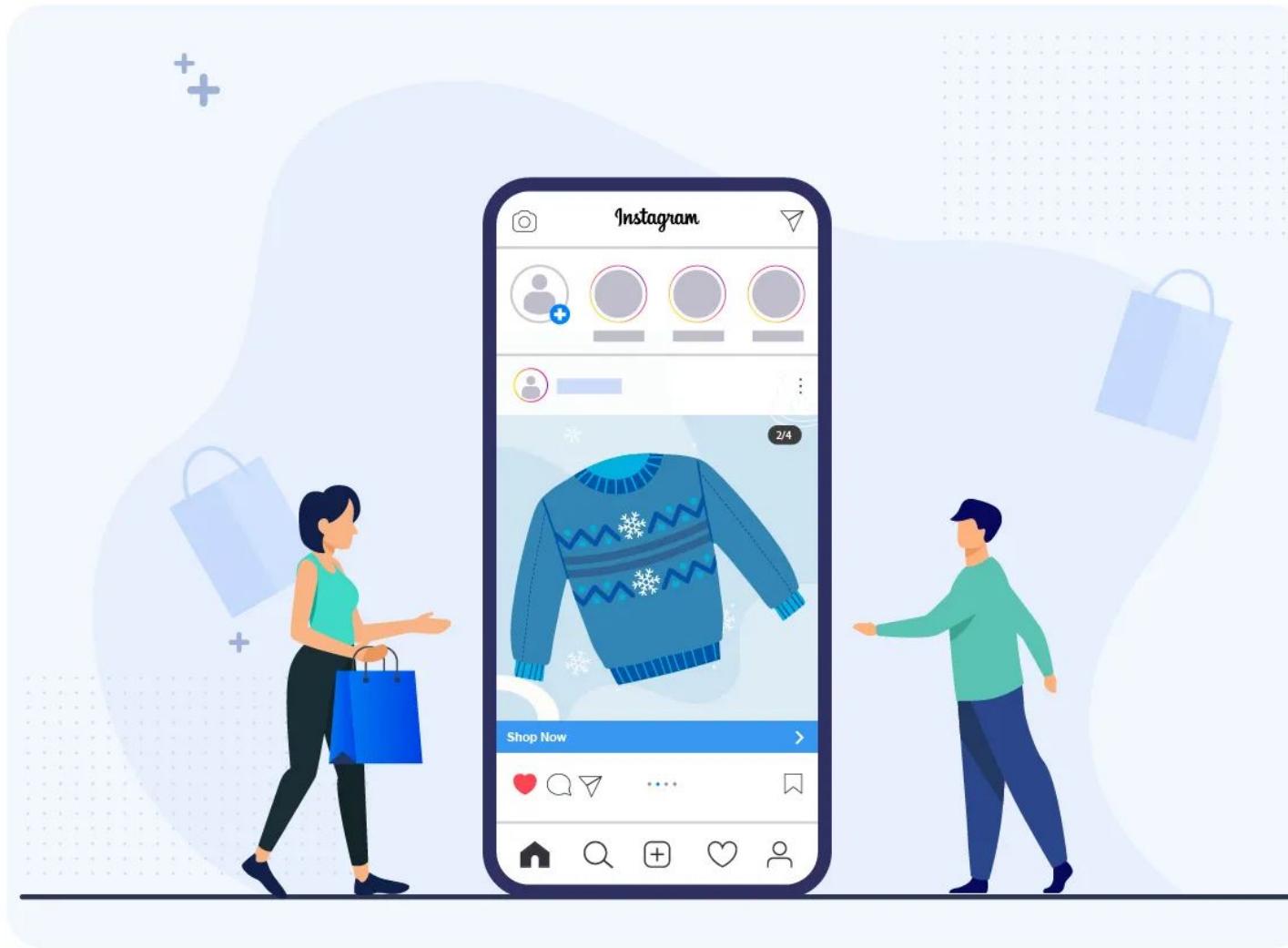


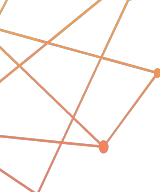
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More...



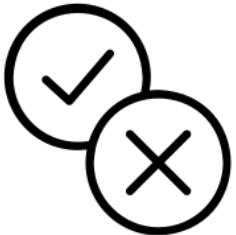


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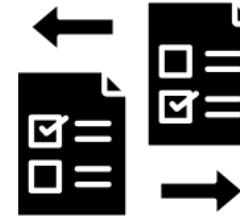


- How “good” recommendations are?
- What is “good”?

Why do we need RS metrics?



Evaluation: They assess RS performance and effectiveness.



Comparison: They help compare different systems or variations.



User Satisfaction: They measure user engagement and they can help in building user trust.



Fairness: They can be used to mitigate biases in recommendations, promoting fairness and inclusivity.



Optimization: They guide improvements and fine-tuning in algorithms and parameters, whereas they support continuous system improvement.



Business Impact: They can be tied to financial outcomes (KPIs).

Science in the background...

Articles About 6,550,000 results (0.13 sec)

Any time [HTML] How good your **recommender system** is? A survey on evaluations in **recommendation** [HTML] springer.com
Since 2023
Since 2022
Since 2019
Custom range...

Sort by relevance [HTML] SemCiR: A citation **recommendation system** based on a novel semantic distance **measure** [HTML] emerald.com
Sort by date

Any type Evaluating **recommendation systems** [PDF] psu.edu
Review articles

 include patents
 include citations

 Create alert

[HTML] T Silveira, M Zhang, X Lin, Y Liu, S Ma - International Journal of Machine ..., 2019 - Springer
... Retention is also a useful metric used in online evaluation of **recommender systems** [32]
user utility and for business. Retention **measures** the impact of the **recommender systems** in ...
☆ Save 99 Cite Cited by 217 Related articles All 3 versions

[HTML] F Zarrinkalam, M Kahani - Program, 2013 - emerald.com
... The purpose of this paper is to propose a novel citation **recommendation system** that inputs a text and recommends publications that should be cited by it. Its goal is to help researchers ...
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[HTML] G Shani, A Gunawardana - **Recommender systems** handbook, 2011 - Springer
... Thus we cannot directly **measure** the **recommender's** influence on user behavior in this setting. Therefore, the goal of the offline experiments is to filter out inappropriate approaches, ...
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[HTML] S Chen, S Owusu, L Zhou - 2013 international conference on ..., 2013 - ieeexplore.ieee.org
... to **measure** the performances of a **recommendation system**. We concluded that the **recommendation system** ... different **measures** towards evaluating a **recommendation system** ...
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Related searches

similarity measure recommendation system	recommendation systems for software engineering
systematic review recommender systems	business opportunities recommendation systems
collaborative filtering recommender systems	good recommendations recommender systems
social network based recommender systems	accuracy metrics recommender systems

Dimensions and metrics for evaluating **recommendation systems** [PDF] github.io



How to measure success?



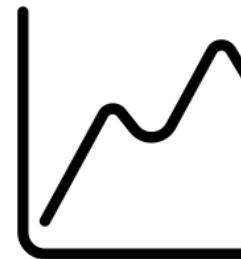
Statistics: quantifies the occurrences of various data entities, such as user interactions, item popularity, or recommendation relevance scores.



Metrics: goes beyond simple counts and offer sophisticated characterizations of Recommender System's performance.



KPIs: Key Performance Indicators (KPIs) focused on business-oriented metrics, which are aligned with the overarching goals of the organization.

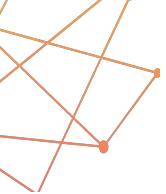


Graphs: visualizations of statistics/metrics across time helping in tracking trends, and identifying seasonality.

What is EOSC?

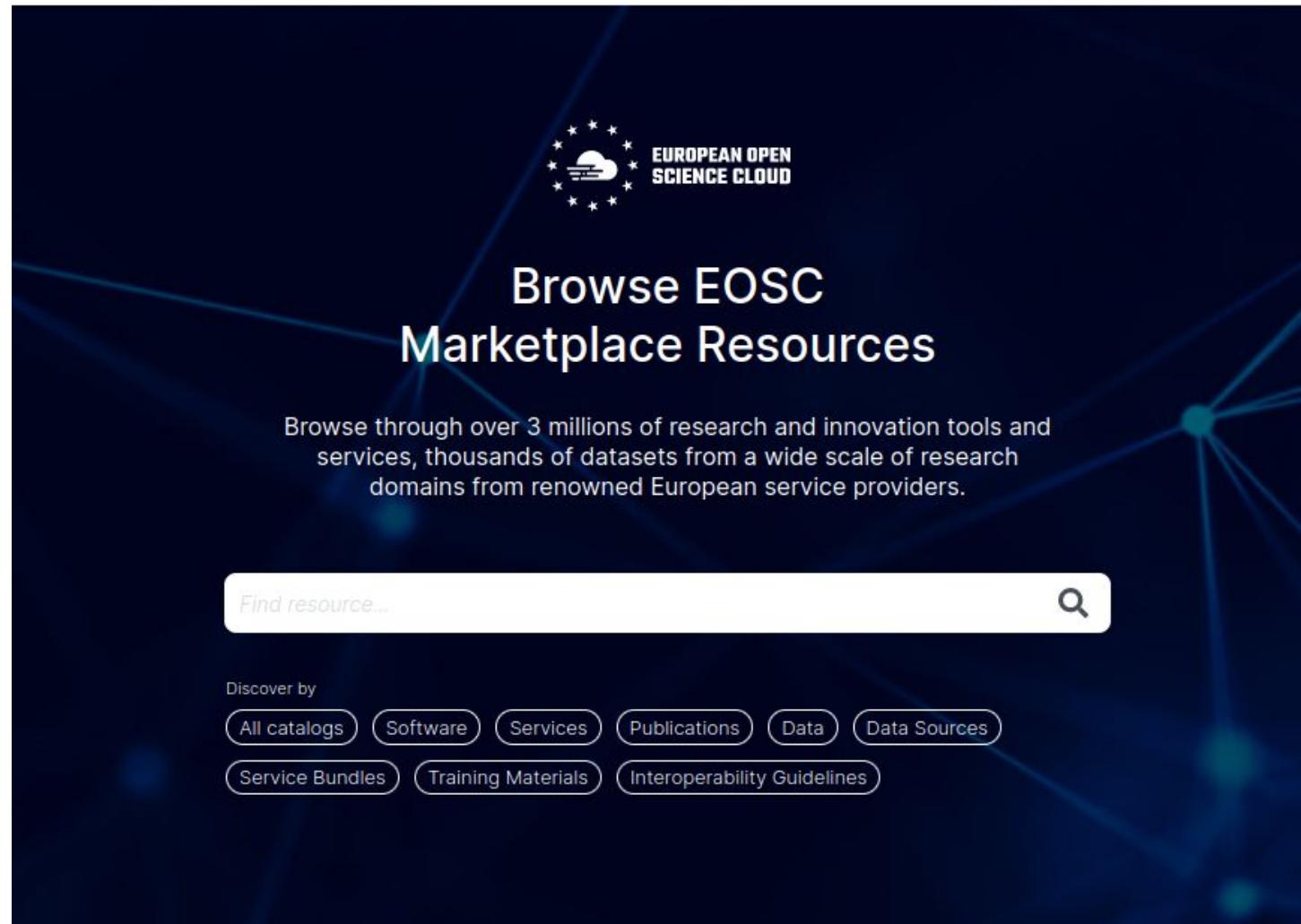
- EOSC stands for the "European Open Science Cloud".
- It is a European initiative aimed at creating a unified, open environment for researchers and scientists.
- Its objectives include enabling access, sharing, and reuse of data, as well as providing access to research services and resources.
- EOSC promotes open science principles, fostering transparency and accessibility in research.
- Collaboration among researchers from various disciplines and institutions is a central goal of EOSC.

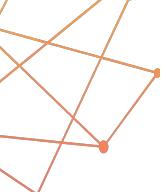




What is EOSC Marketplace?

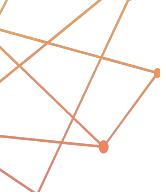
- The EOSC Marketplace is an online platform and ecosystem within EOSC: <https://marketplace.eosc-portal.eu/>
- It is designed to facilitate access to a wide range of digital resources, services, and data.
- It serves as a centralized hub for researchers and scientists in the EU research community.
- Researchers can use the marketplace to discover, access, and use resources that support their research activities.
- Resources available through the marketplace include data repositories, software applications, computing and storage facilities, trainings, and various research services.





What is Recommender Metrics Framework (RMF)?

- A Python open-source software that monitors, analyzes, and evaluates recommendation mechanisms.
- Measures the effectiveness of the EOSC Marketplace RS to enhance the user experience and improve AI algorithms.
- Incorporates diagnostic statistics, metrics and visualizations for deeper insights into model performance.
- Presents reports as a web service and visualizes statistics, metrics, and Key Performance Indicators (KPIs) through a RESTful API and UI dashboard.
- Quantitative evaluation is taking into account EOSC Marketplace resources, user actions, and recommendations.
- Supports real-time and offline data ingestion, multiple resource types, and various recommendation engines as sources.
- Evolves over time, adding features and utilities to promote the development of more reliable and high-quality RS designs.



Use Cases

IS CURRENTLY USED

CAN BE USED

Monitoring the EOSC Marketplace RS

- Monitors and reports diagnostic metrics for the EOSC Marketplace RS.
- Analyses user actions and recommendations.
- Provides Statistics, Metrics, KPIs, Graphs in a REST API and dashboard UI.
- Delivers comprehensive documentation.

Evaluate a third-party RS

- An analysis tool of the recommendation engine.
- Data preparation with the necessary input information.
- Tasks involve retrieving data from multiple sources, removing irrelevant data, correlating information, and generating **statistical insights**.

What it offers?

Statistics

- Number of Users
- Number of Resources
- Number of Recommended items
- Number of User Actions by
 - Registered or
 - Anonymous users
- Total Views

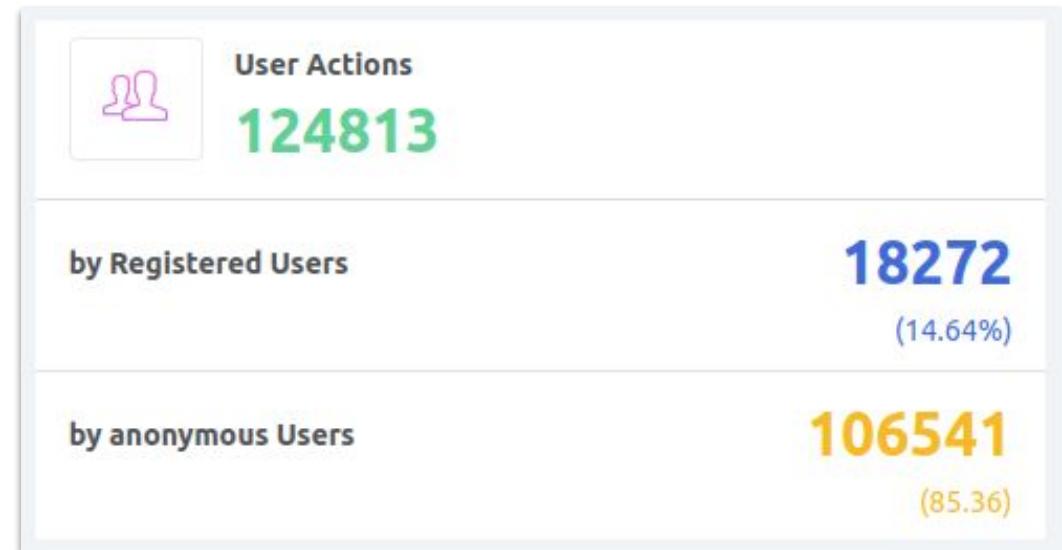


What it offers?

User Actions

Statistics

- Number of Users
- Number of Resources
- Number of Recommended items
- Number of User Actions by
 - Registered or
 - Anonymous users
- Total Views



What it offers?

Metrics

- Accuracy
- Catalog Coverage
- Diversity Gini Index
- Diversity Shannon Entropy
- Novelty
- User Coverage

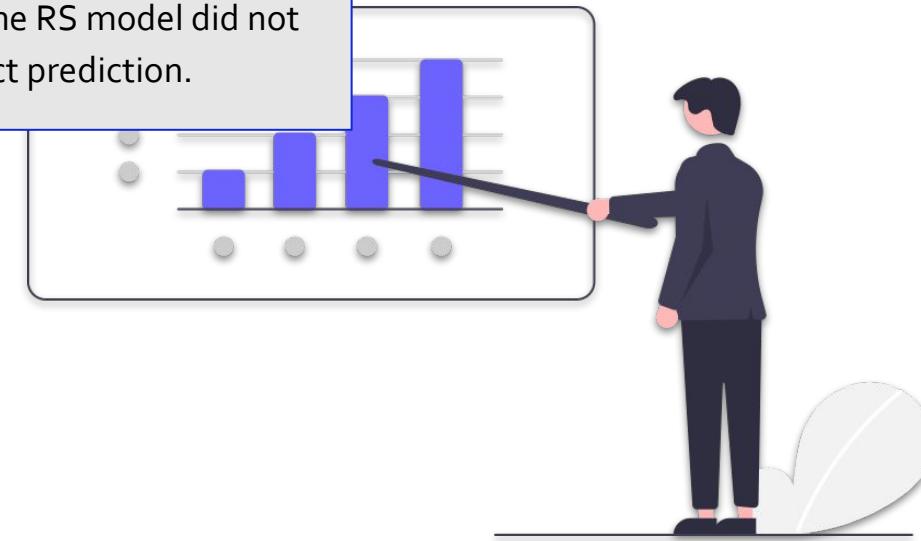


What it offers?

Metrics

- Accuracy
- Catalog Coverage
- Diversity Gini Index
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- User Coverage

Measures Recommendations' accuracy based on users' access to the services. A value of 1, indicates that the RS model got all the predictions right, and a value of 0 indicates that the RS model did not make a single correct prediction.

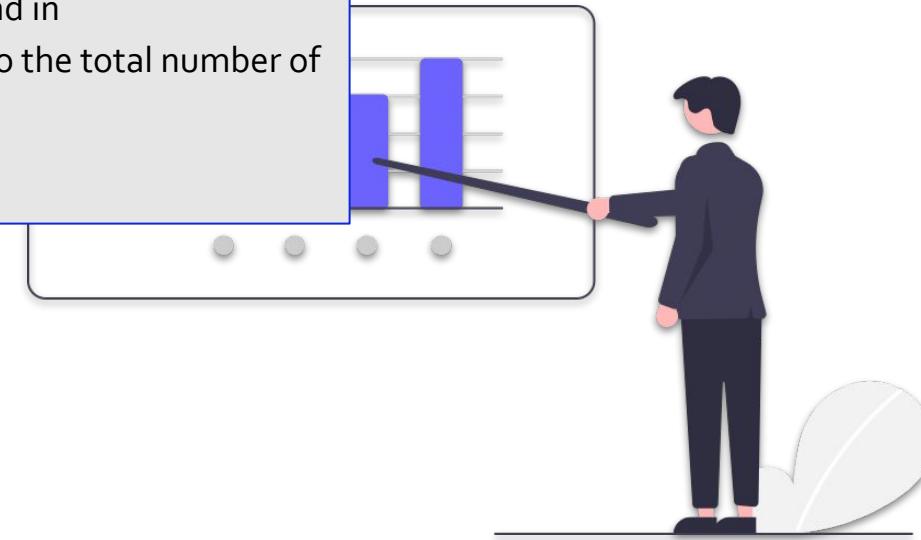


What it offers?

Metrics

- Accuracy
- **Catalog Coverage**
- Diversity Gini Index
- Diversity Shannon Entropy
- Novelty
- User Coverage

The percentage (%) of the division of the unique services found in recommendations to the total number of published services.



What it offers?

Metrics

- Accuracy
- Catalog Coverage
- **Diversity Gini Index**
- Diversity Shannon Entropy
- Novelty
- User Coverage

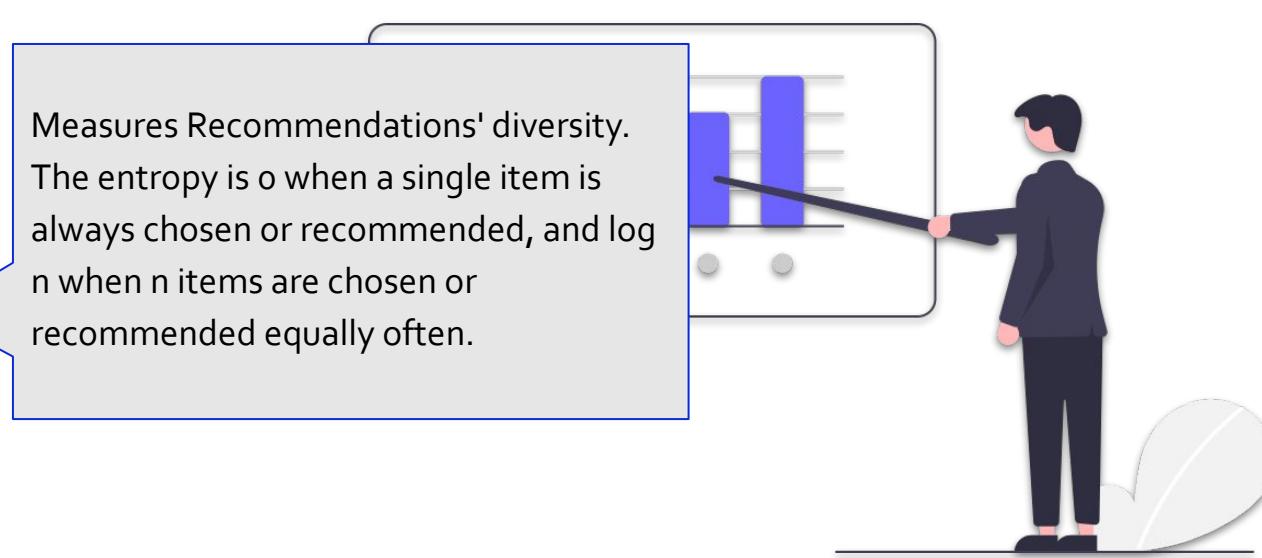
Measures Recommendations' diversity.
The index is 0 when all items are chosen
equally often, and 1 when a single item is
always chosen.



What it offers?

Metrics

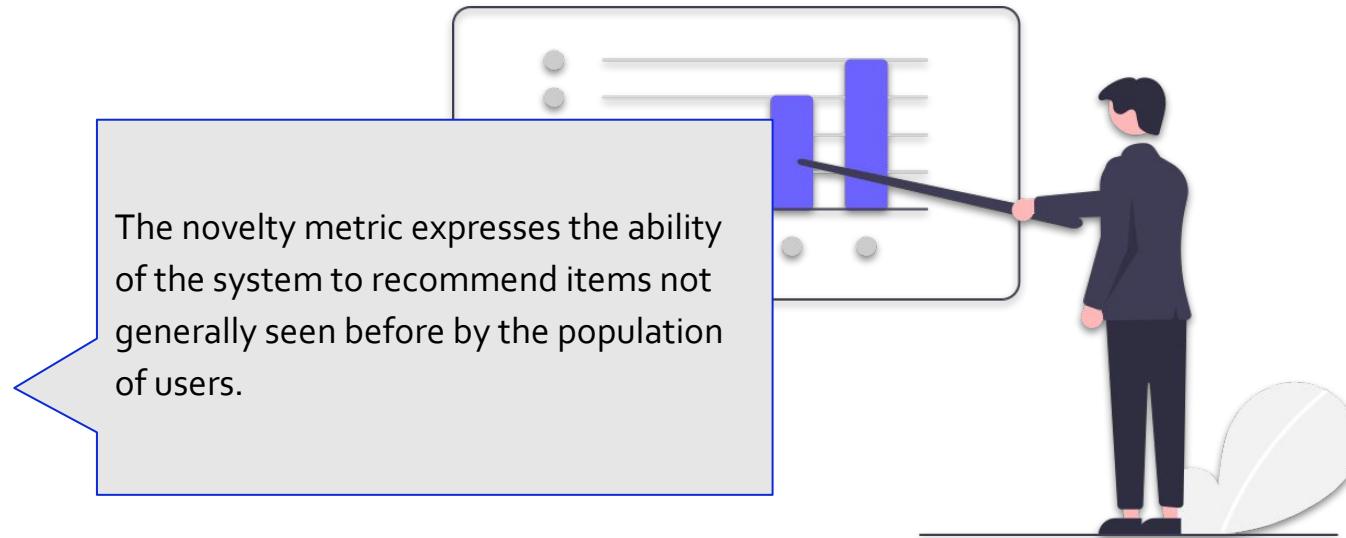
- Accuracy
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- **Diversity Shannon Entropy**
- Novelty
- User Coverage



What it offers?

Metrics

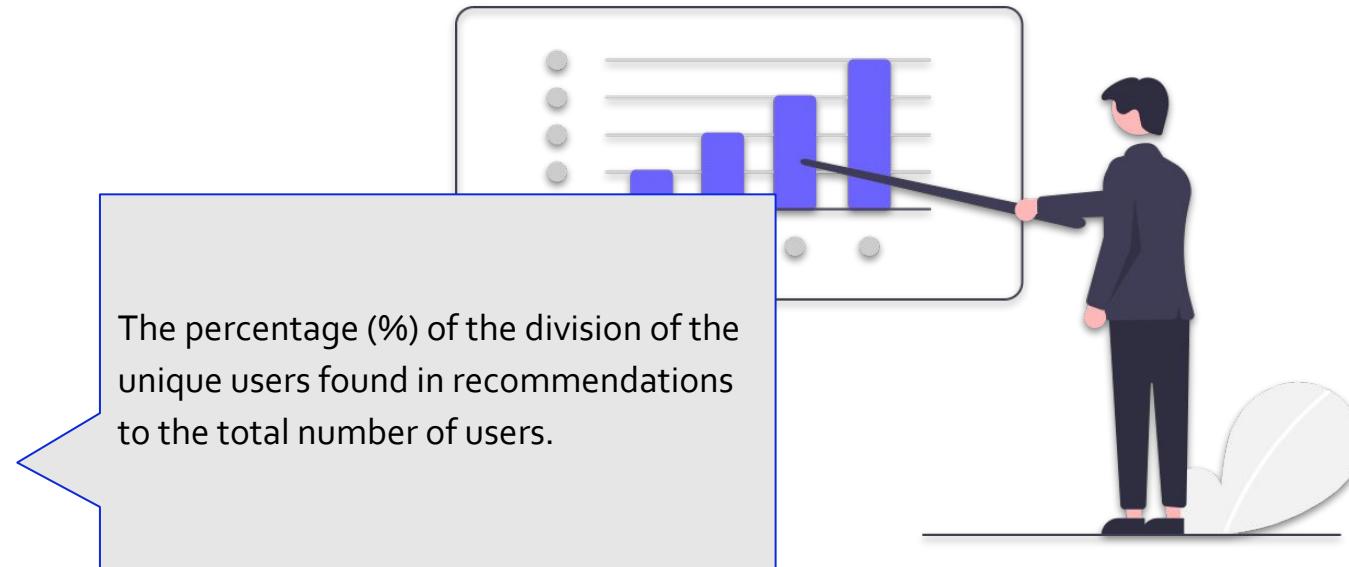
- Accuracy
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- **Novelty**
- User Coverage



What it offers?

Metrics

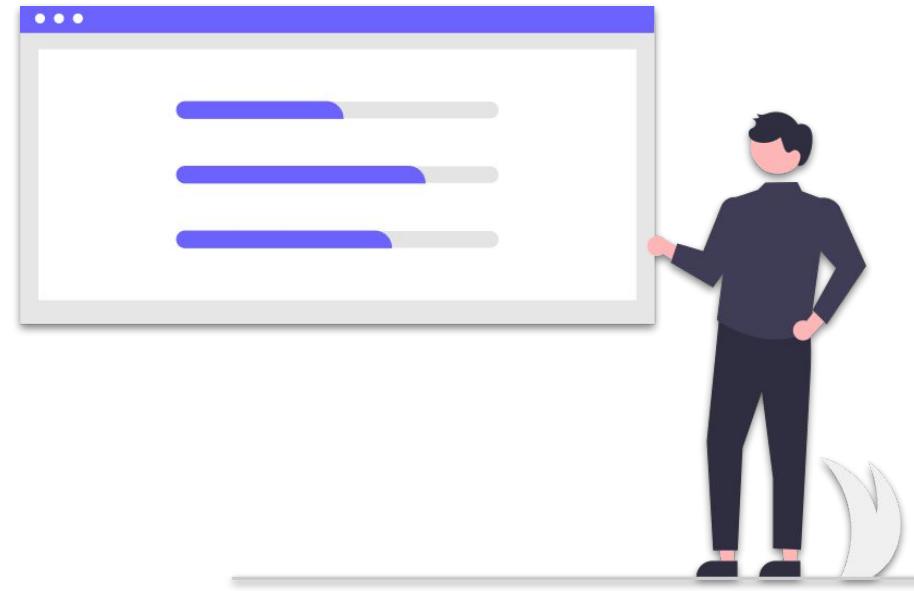
- Accuracy
- Catalog Coverage
- Diversity Gini Index
- Diversity Shannon Entropy
- Novelty
- **User Coverage**



What it offers?

KPIs

- Click-Through Rate
- Hit-Rate
- Top 5 viewed Items
- Top 5 recommended Items
- Top 5 viewed categories
- Top 5 recommended categories
- Top 5 viewed scientific domains
- Top 5 recommended scientific domains

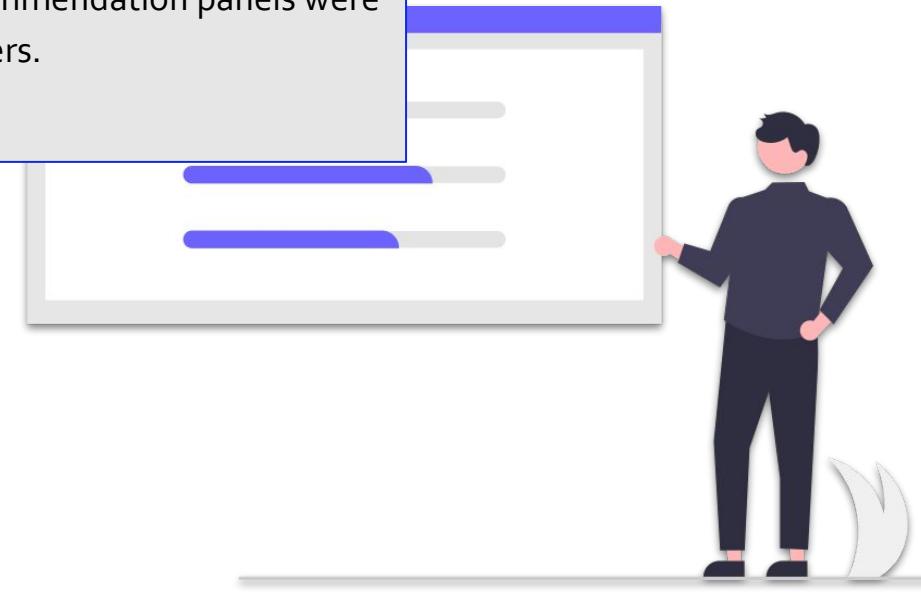


What it offers?

KPIs

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- Top 5 viewed Items
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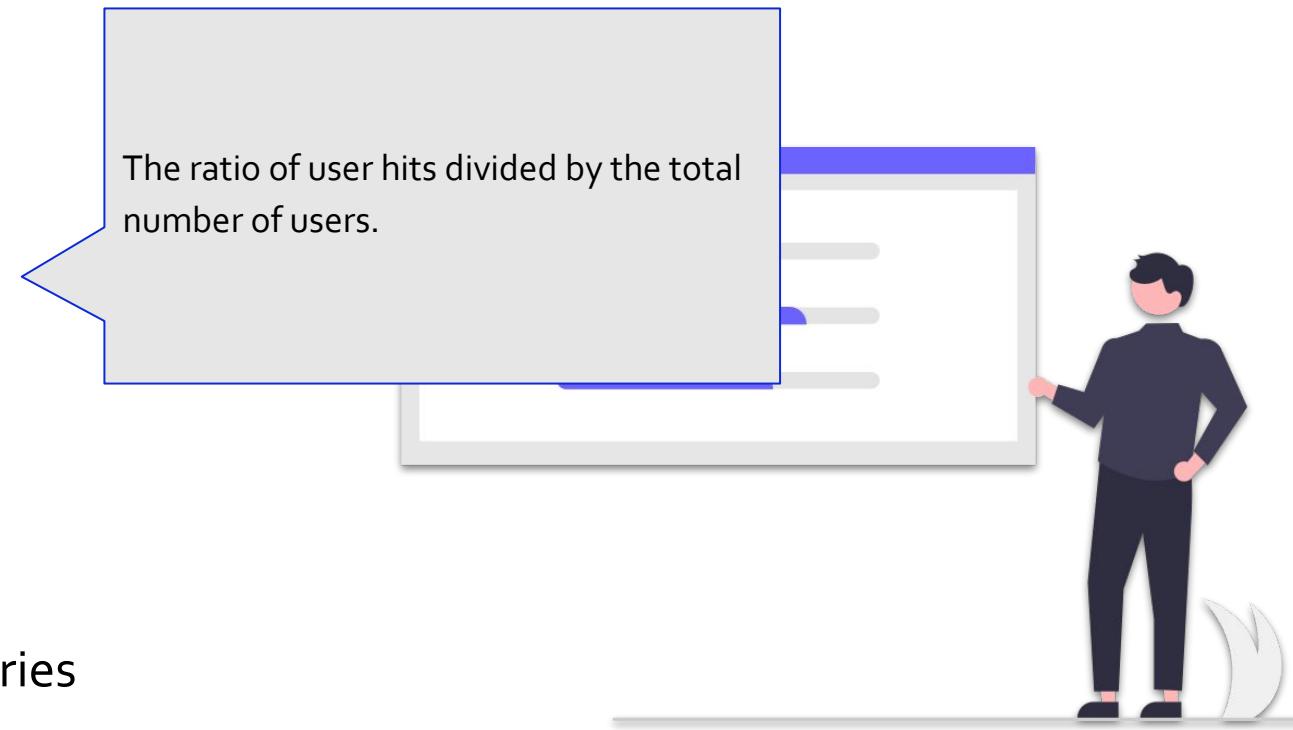
The number of user clicks through recommendations panels divided by the total times recommendation panels were presented to users.



What it offers?

KPIs

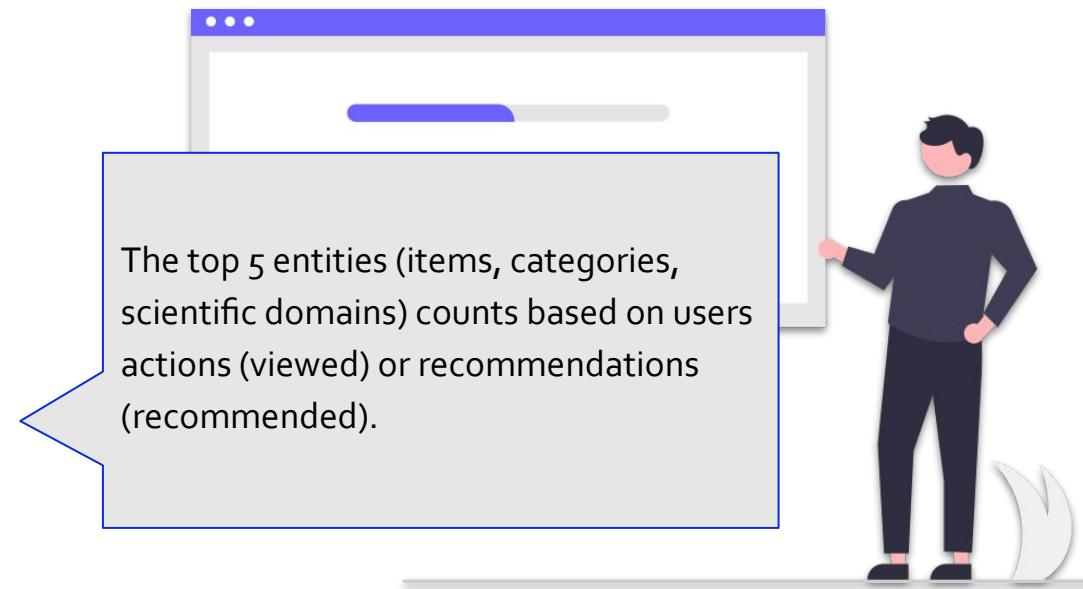
- Click-Through Rate
- **Hit-Rate**
- Top 5 viewed Items
- Top 5 recommended Items
- Top 5 viewed categories
- Top 5 recommended categories
- Top 5 viewed scientific domains
- Top 5 recommended scientific domains



What it offers?

KPIs

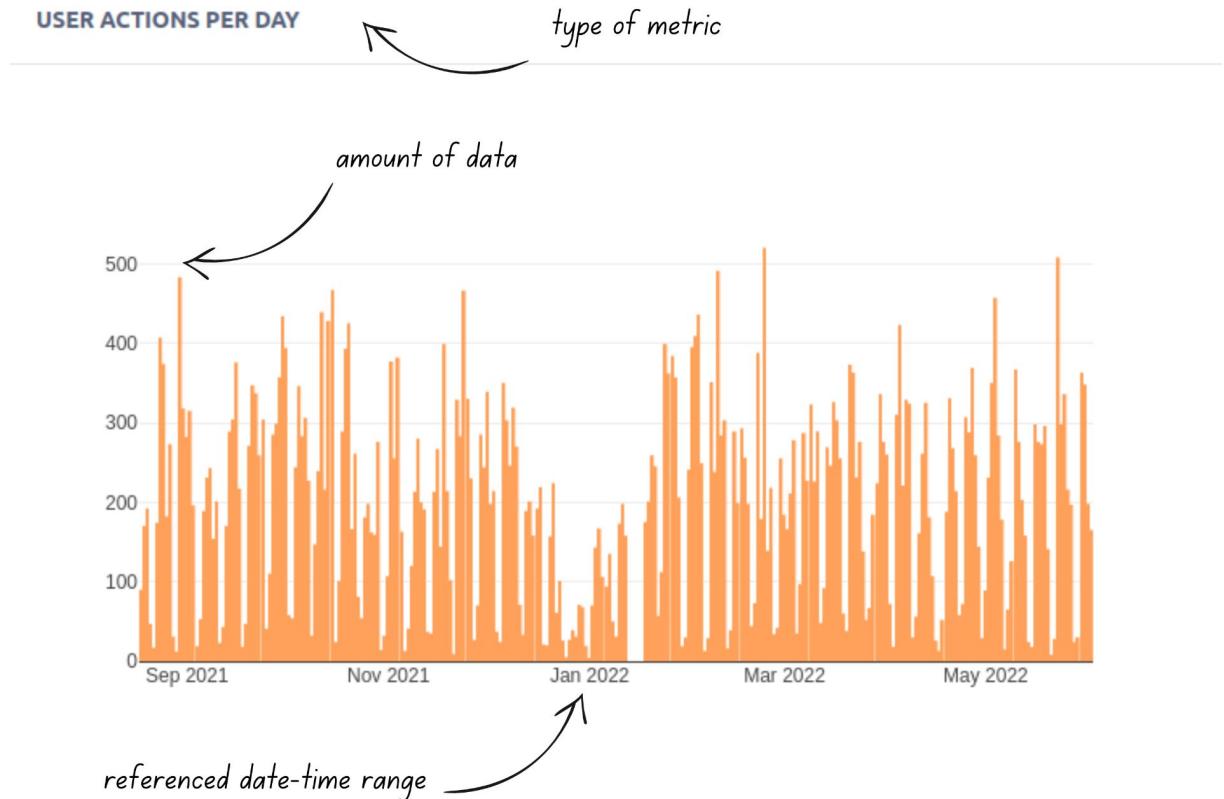
- Click-Through Rate
- Hit-Rate
- **Top 5 viewed Items**
- **Top 5 recommended Items**
- **Top 5 viewed categories**
- **Top 5 recommended categories**
- **Top 5 viewed scientific domains**
- **Top 5 recommended scientific domains**



What it offers?

Graphs

- User Actions per day
- Recommended Items per day
- User Actions per month
- Recommended Items per month



What it offers?

Rest API

- Statistics
- Metrics
- KPIs
- Graphs' Data

```
▼ 0:
    name: "accuracy"
    value: 0.9942
    ▼ doc:
        "The mean value of the accuracy score found for each user defined by the fraction of correct recommendations given to the user"

▼ 1:
    name: "catalog_coverage"
    value: 82.43
    ▼ doc:
        "The percentage (%) of unique services to the total number of services"

▼ 2:
    name: "click_through_rate"
    value: 0.03
    ▼ doc:
        "The number of user clicks through recommendations panels divided by the total times the user has interacted with the recommendation system"

▼ 3:
    name: "diversity"
    value: 3.2668
    ▼ doc:
        "The diversity of the recommendations according to Shannon Entropy. The entropy is 0 when all recommendations are the same and reaches its maximum when all recommendations are different"

▼ 4:
    name: "diversity_gini"
    value: 0.9718
    ▼ doc:
        "The diversity of the recommendations according to GiniIndex. The index is 0 when all recommendations are the same and reaches its maximum when all recommendations are different"

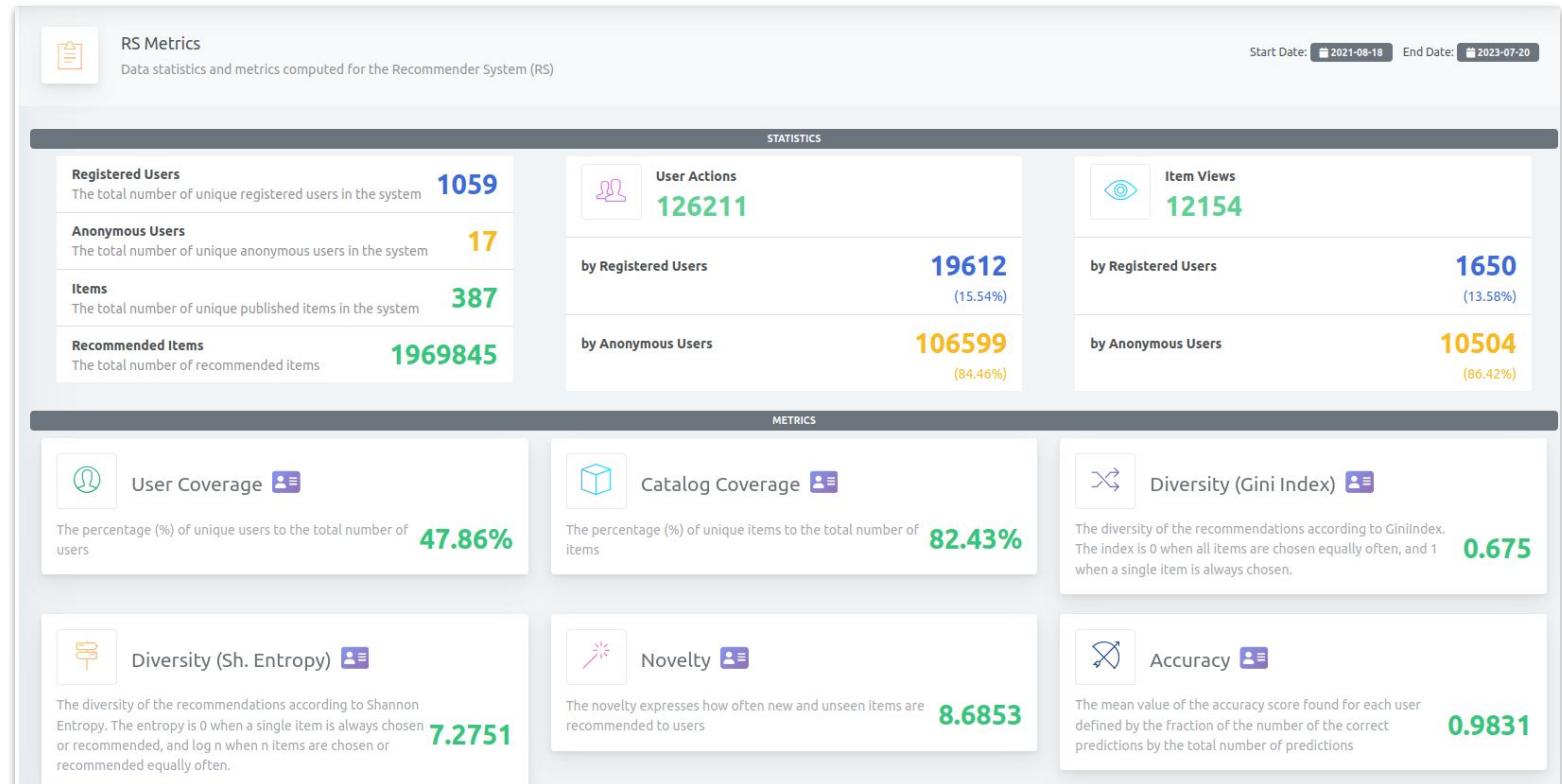
▼ 5:
    name: "hit_rate"
    value: 0.01275
    ▼ doc:
        "The ratio of user hits divided by the total number of users (user hit: a user that has interacted with at least one recommended item)"

▼ 6:
    name: "novelty"
    value: 8.0964
    ▼ doc:
        "The novelty expresses how often new and unseen items are recommended to users"
```

What it offers?

UI Dashboard

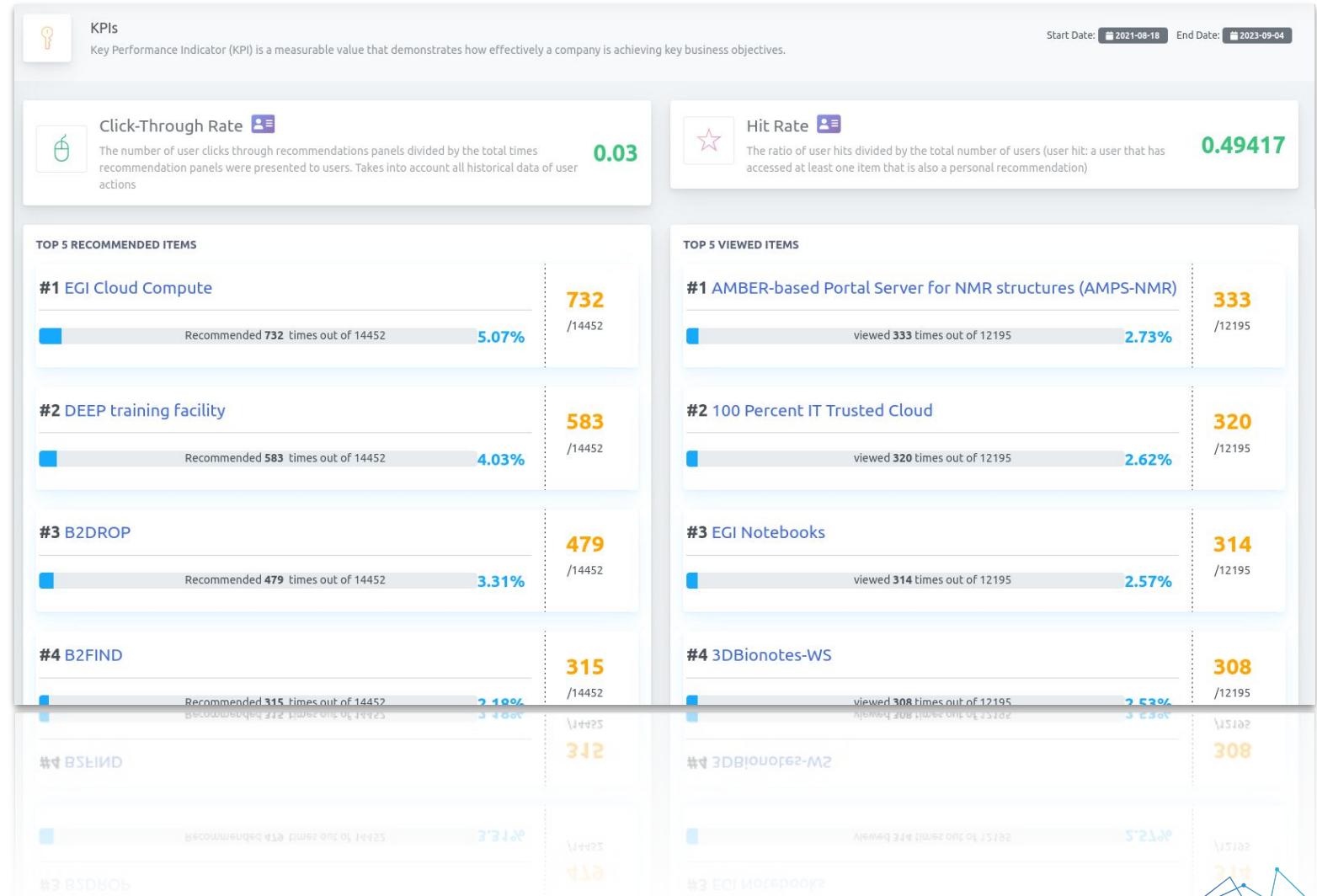
- Statistics
- Metrics
- KPIs
- Graphs



What it offers?

UI Dashboard

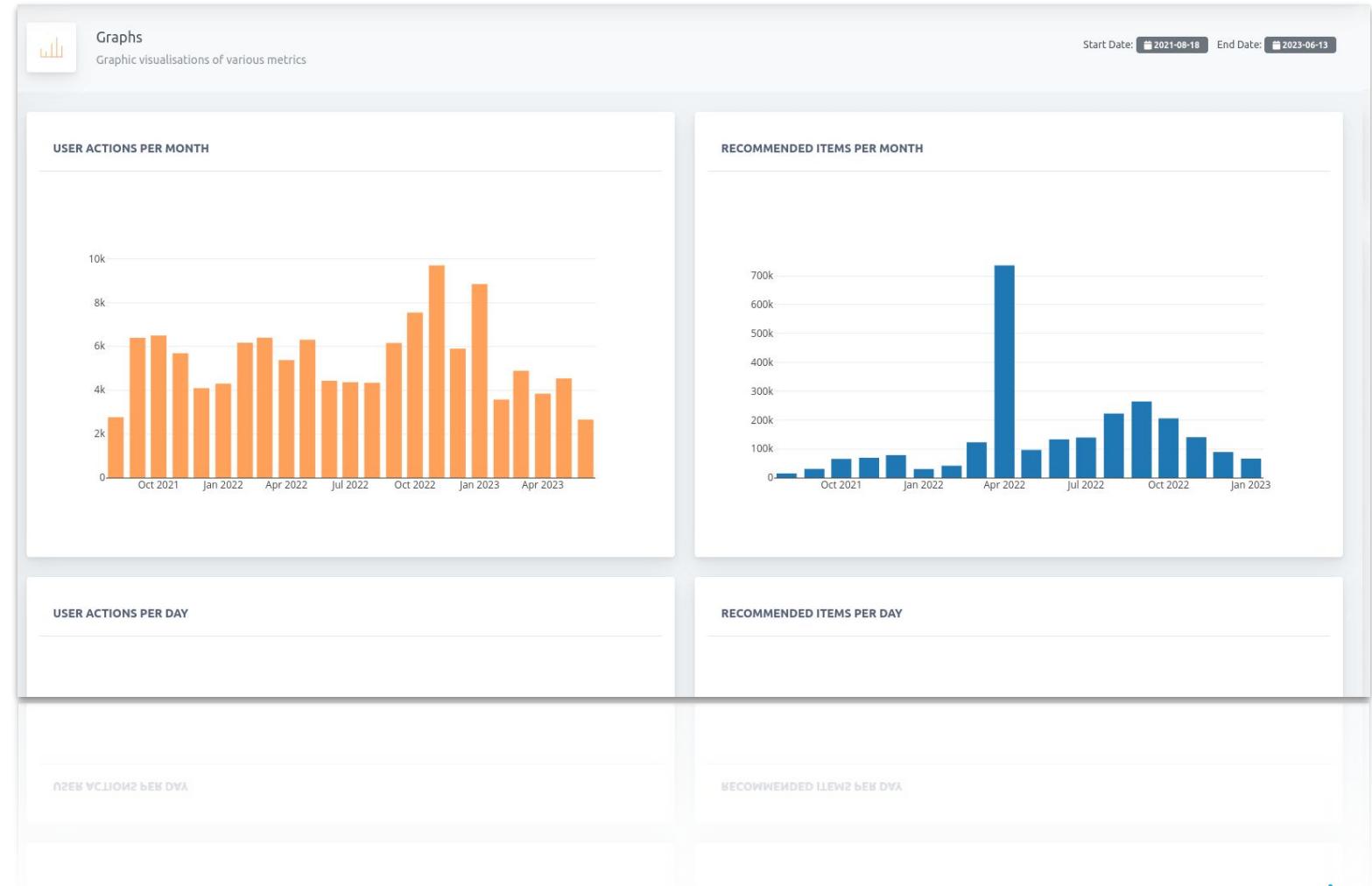
- Statistics
- Metrics
- KPIs
- Graphs



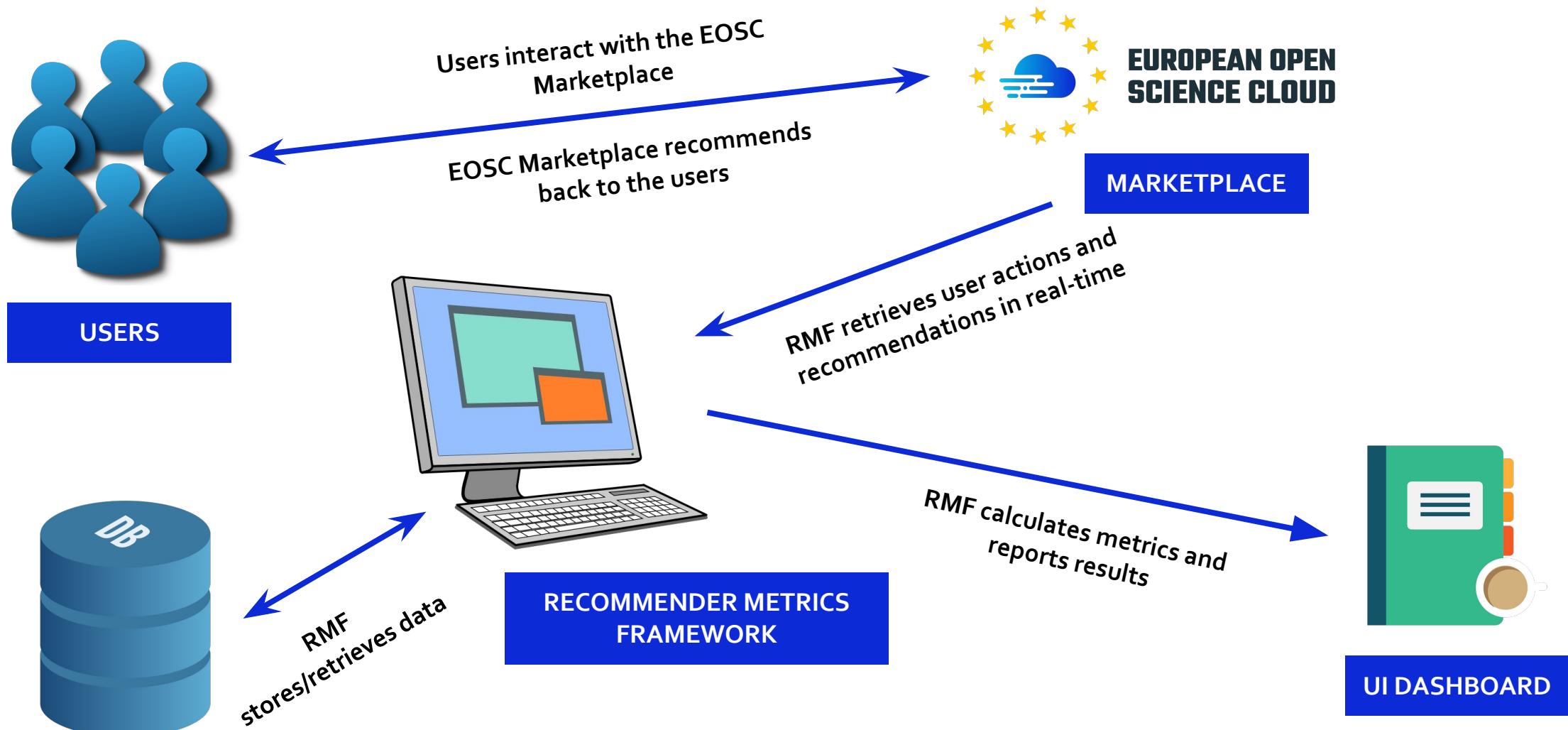
What it offers?

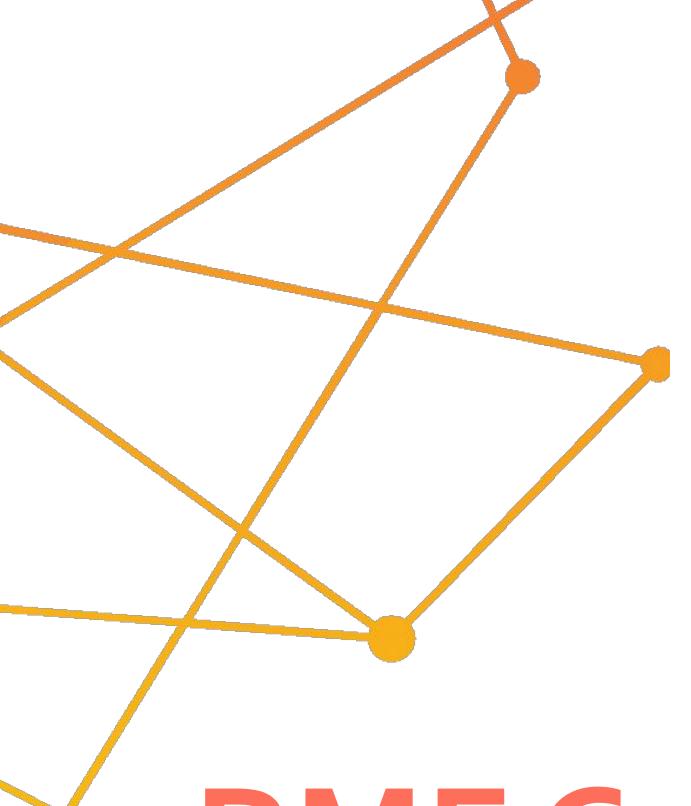
UI Dashboard

- Statistics
- Metrics
- KPIs
- Graphs



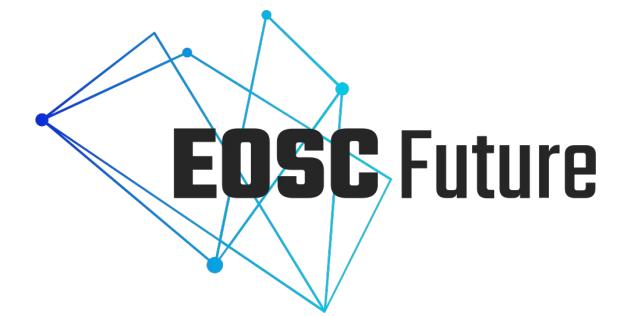
Process Flow



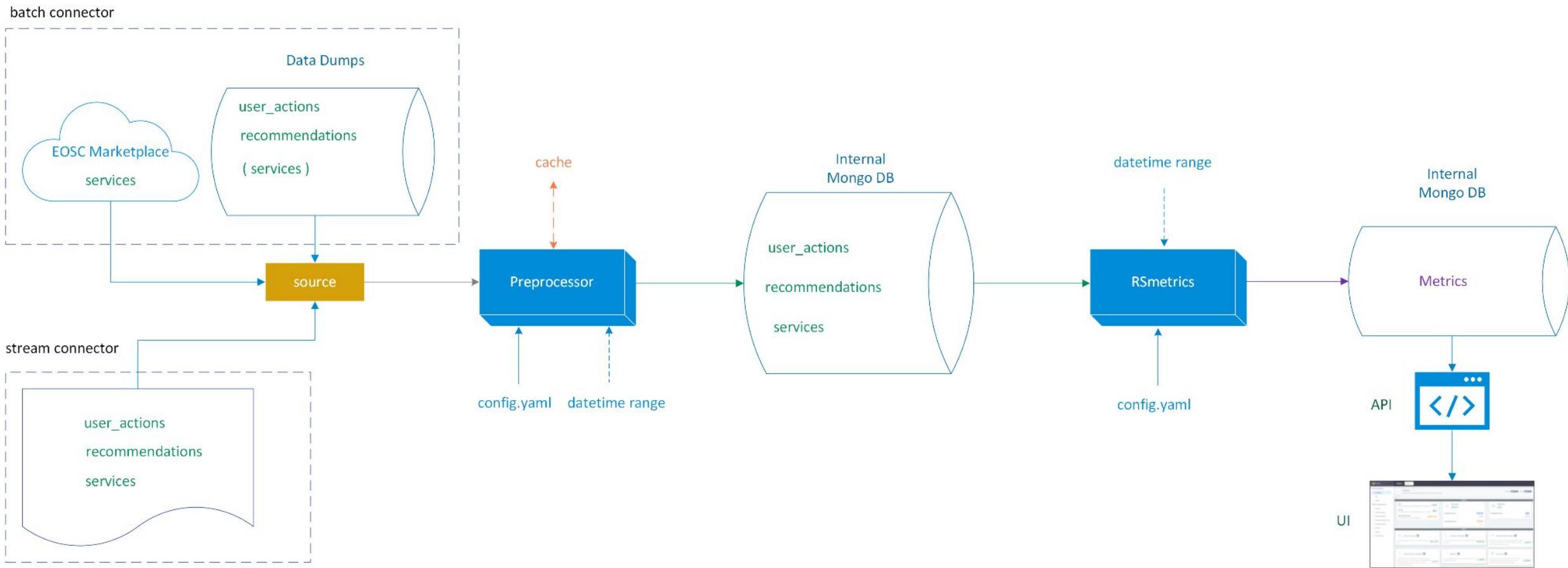


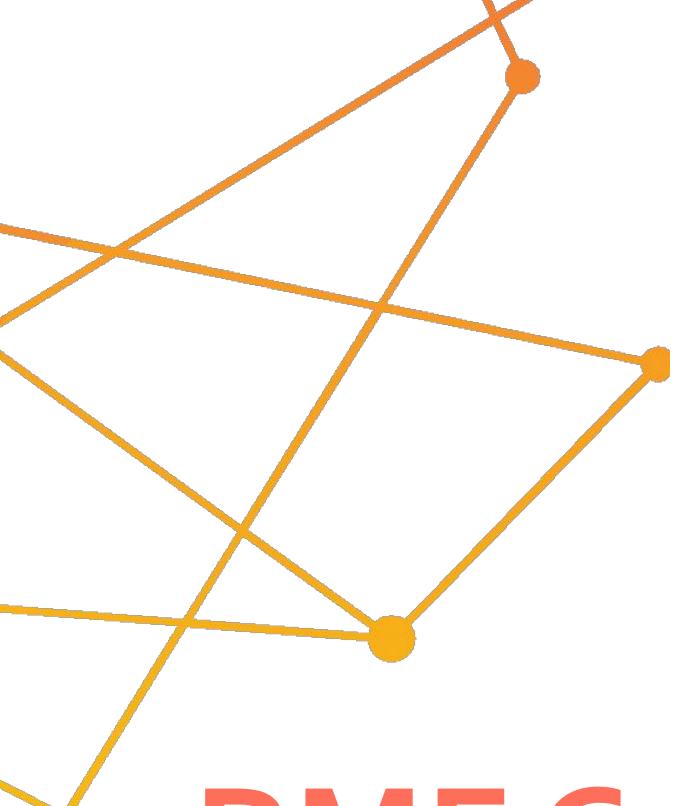
RMF Components

All Units



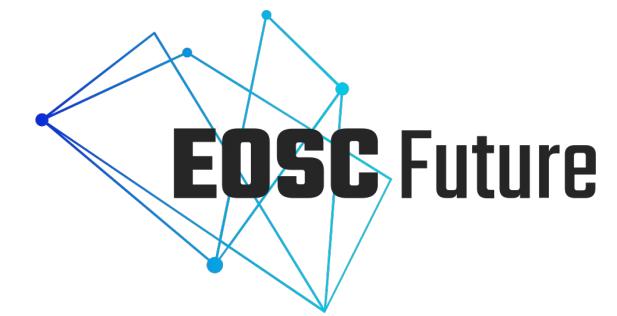
Framework's components



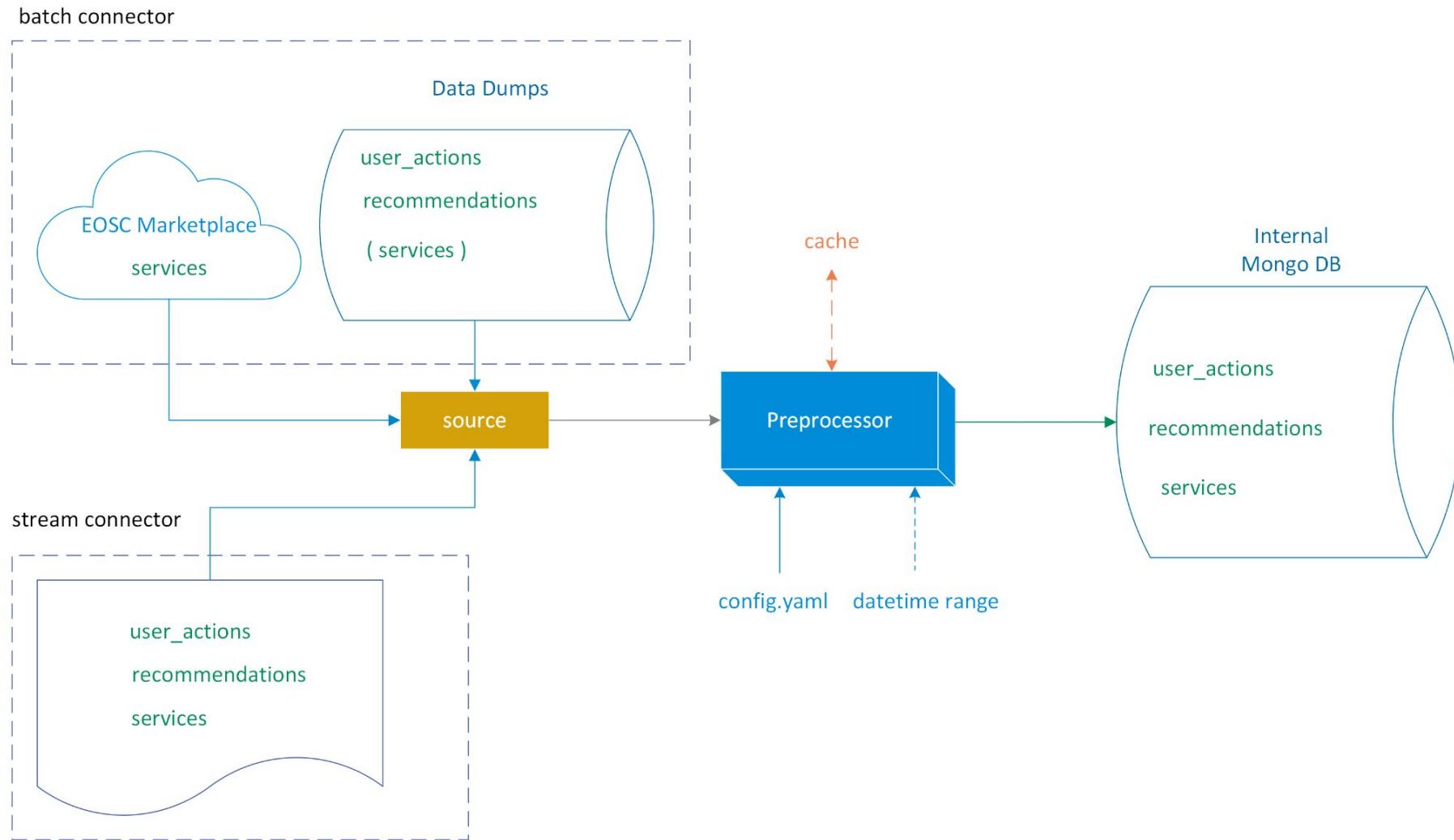


RMF Components

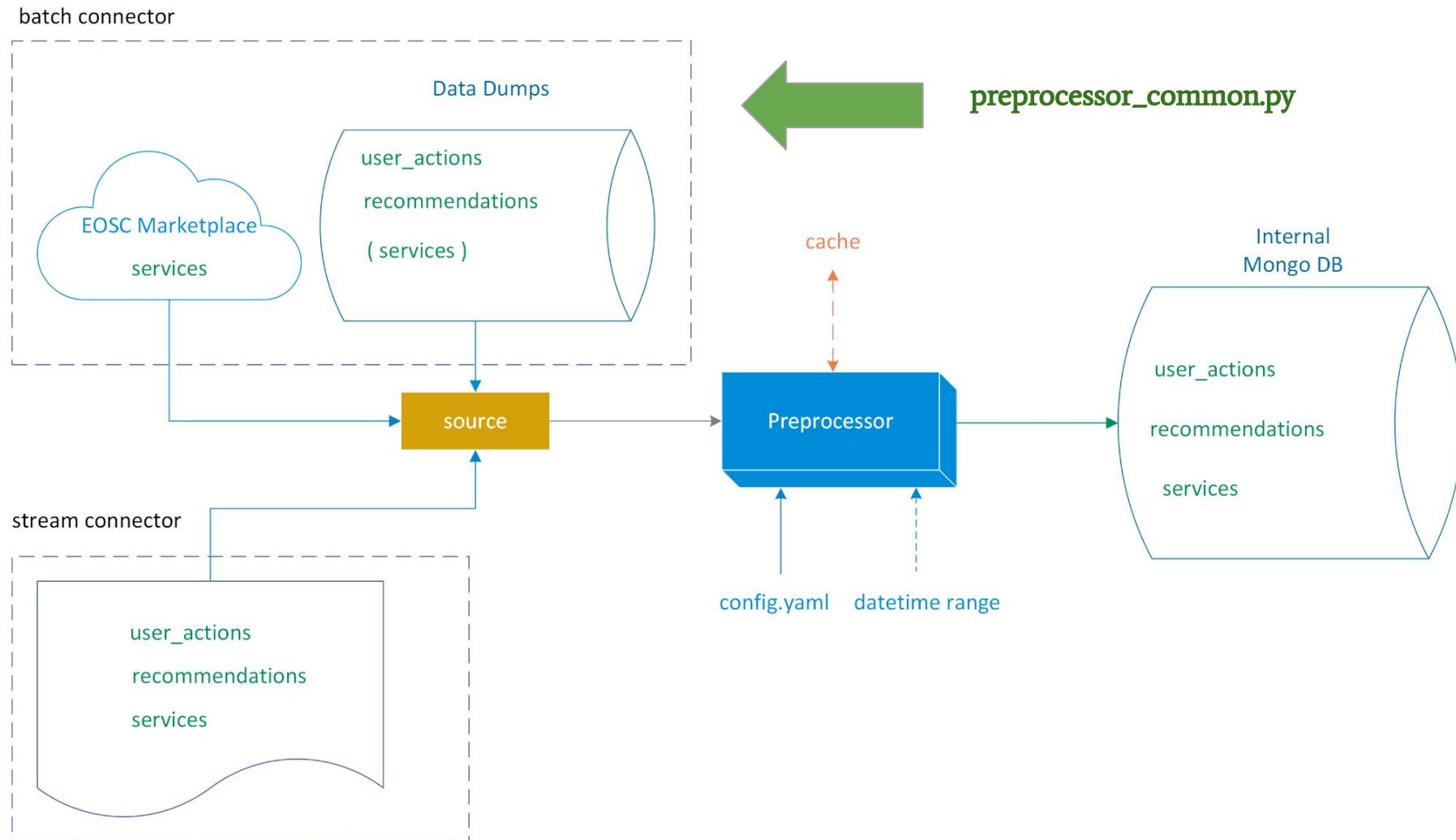
Preprocessor Unit



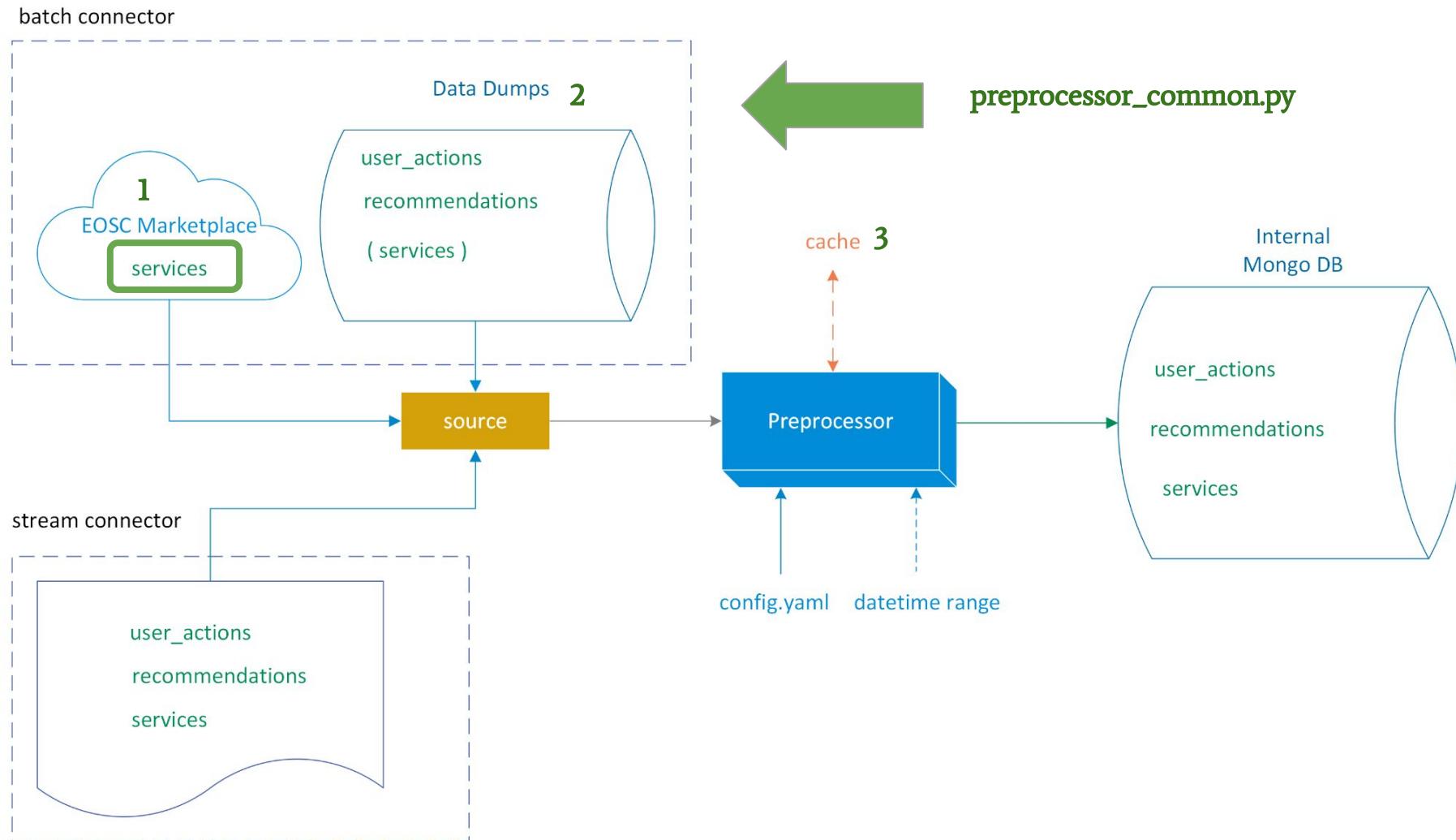
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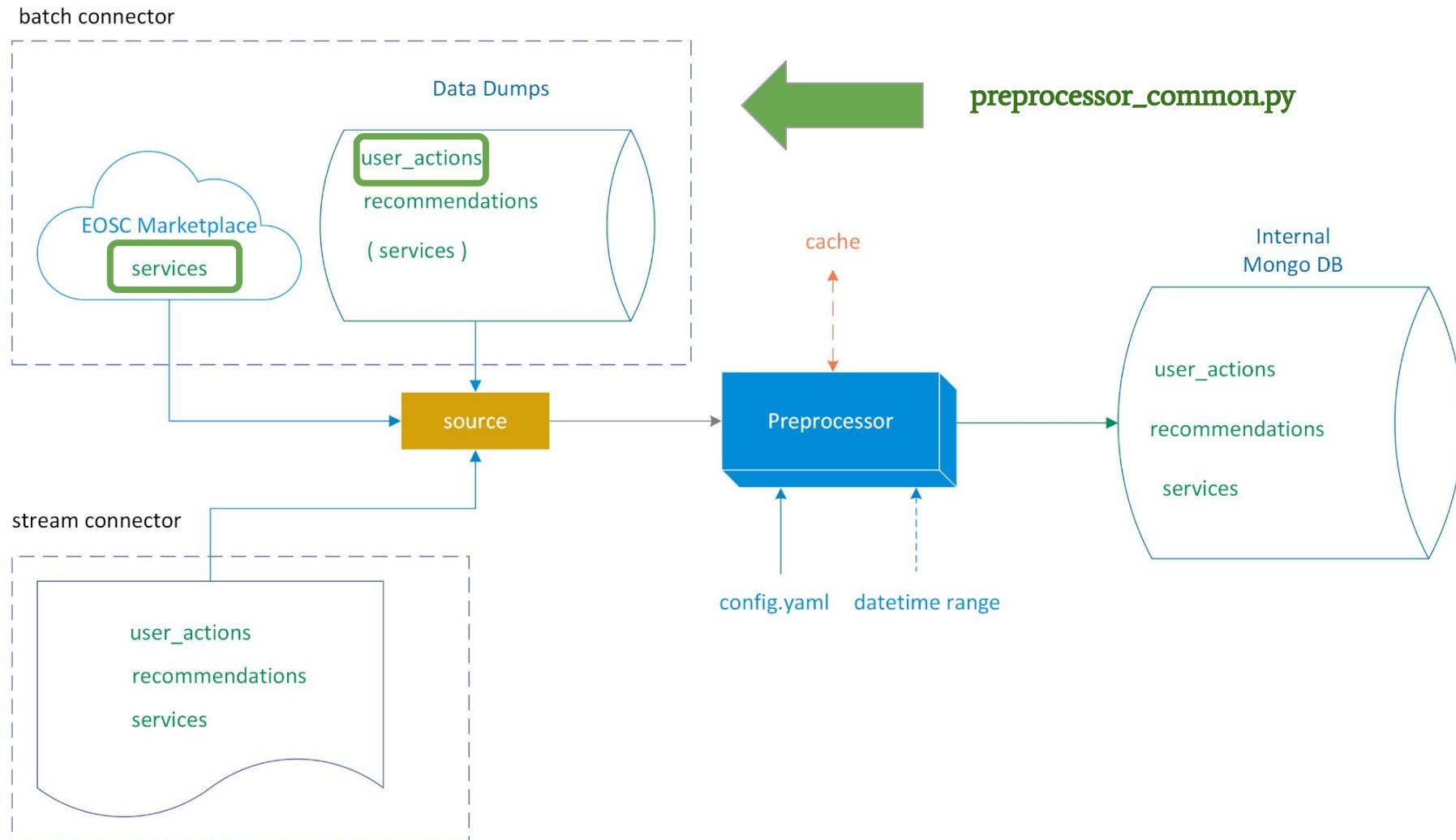
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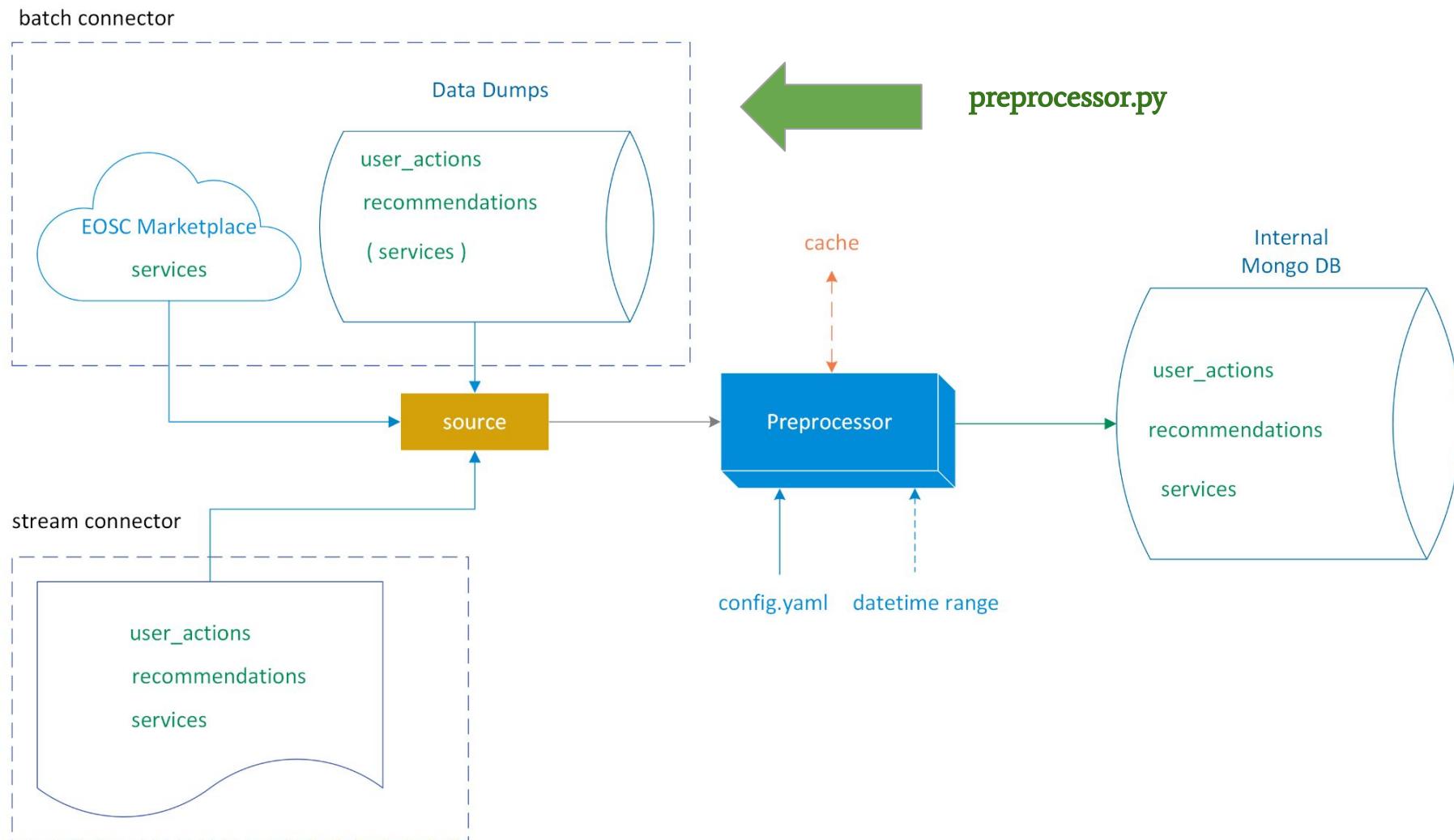
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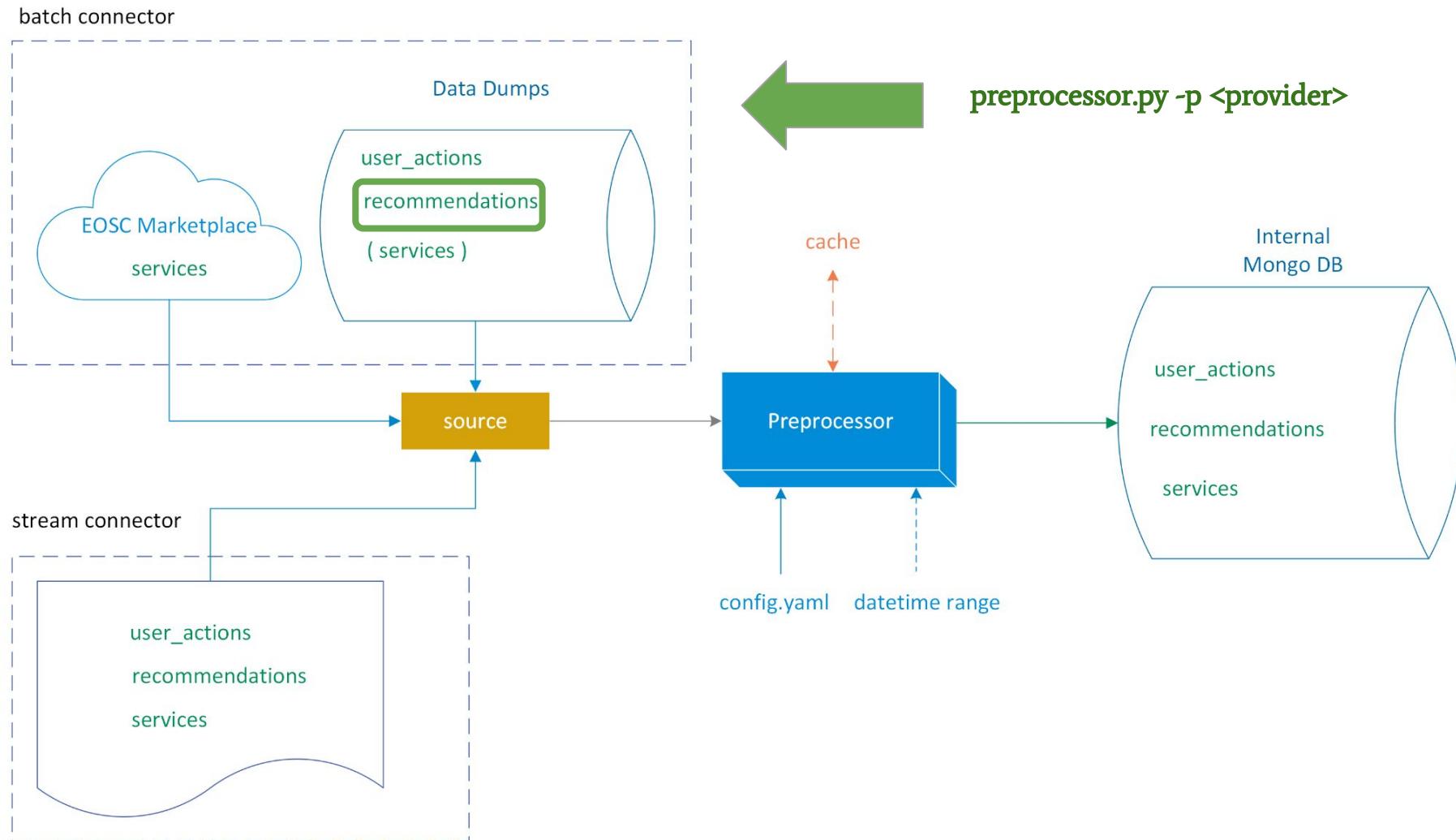
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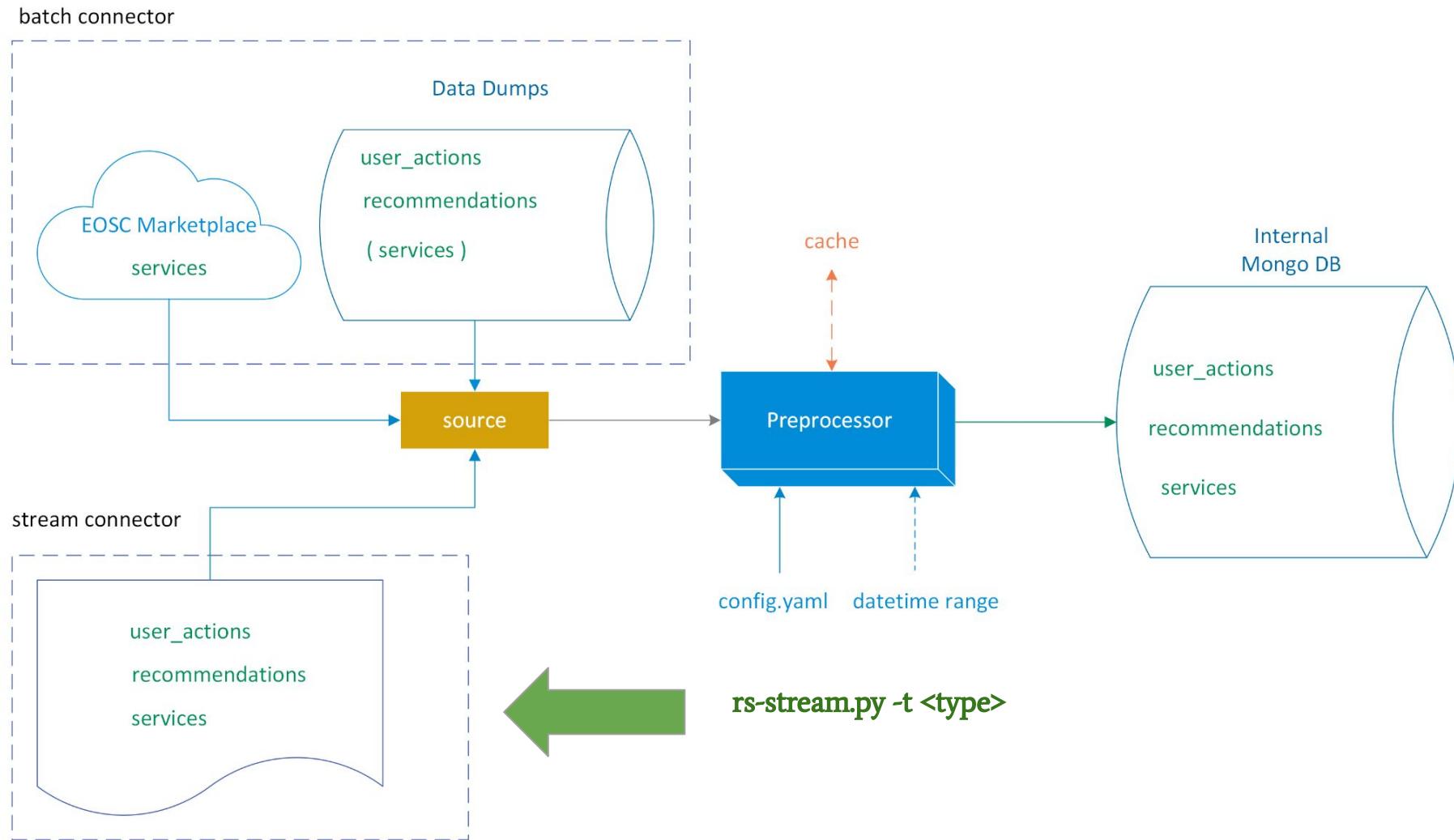
Preprocessor Unit

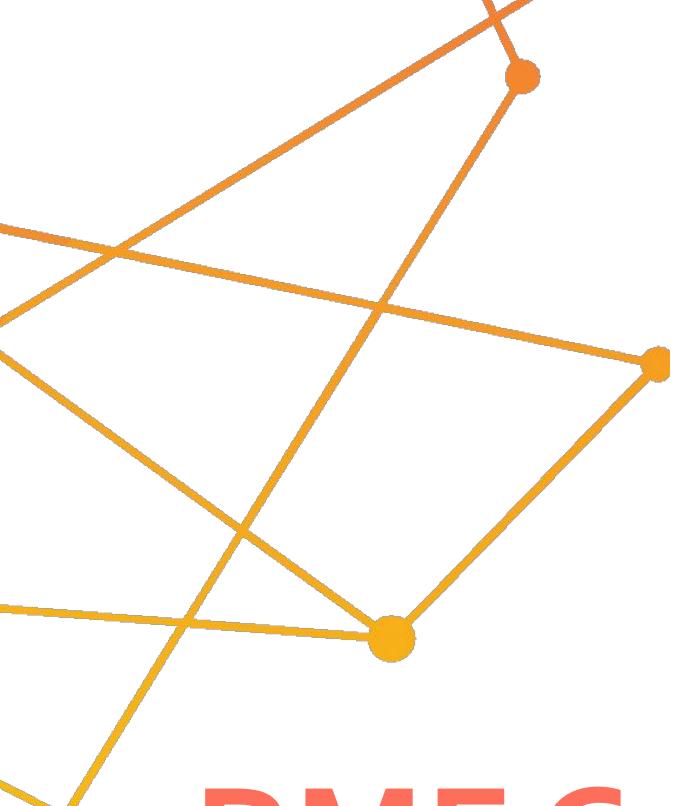


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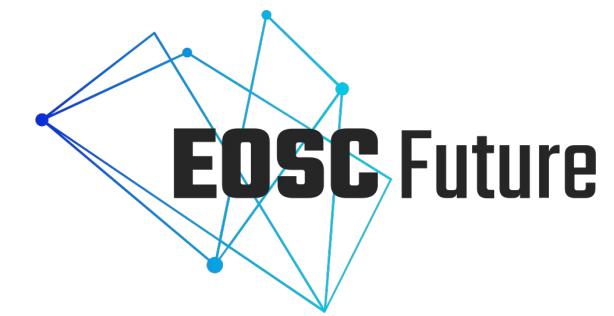
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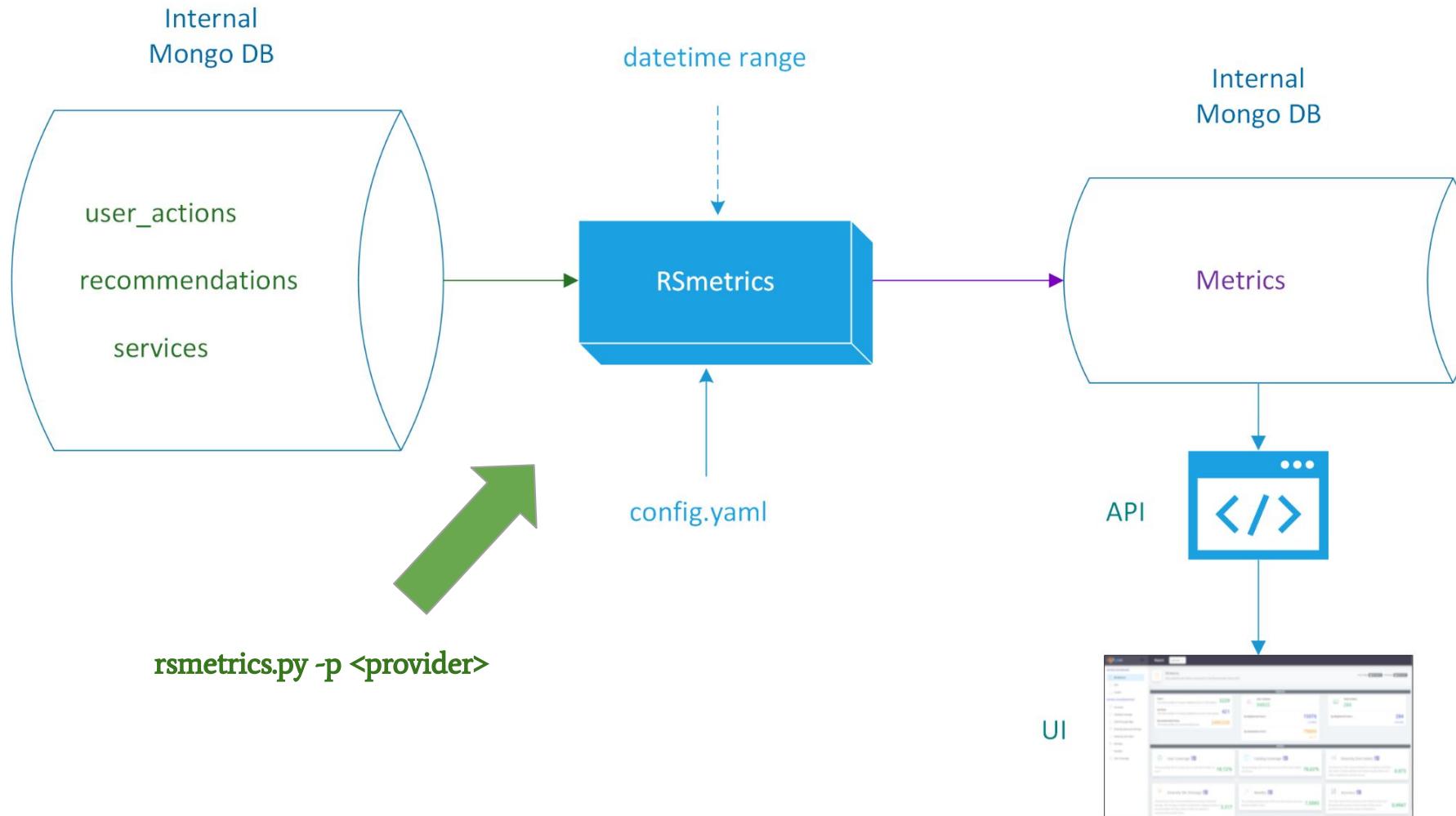


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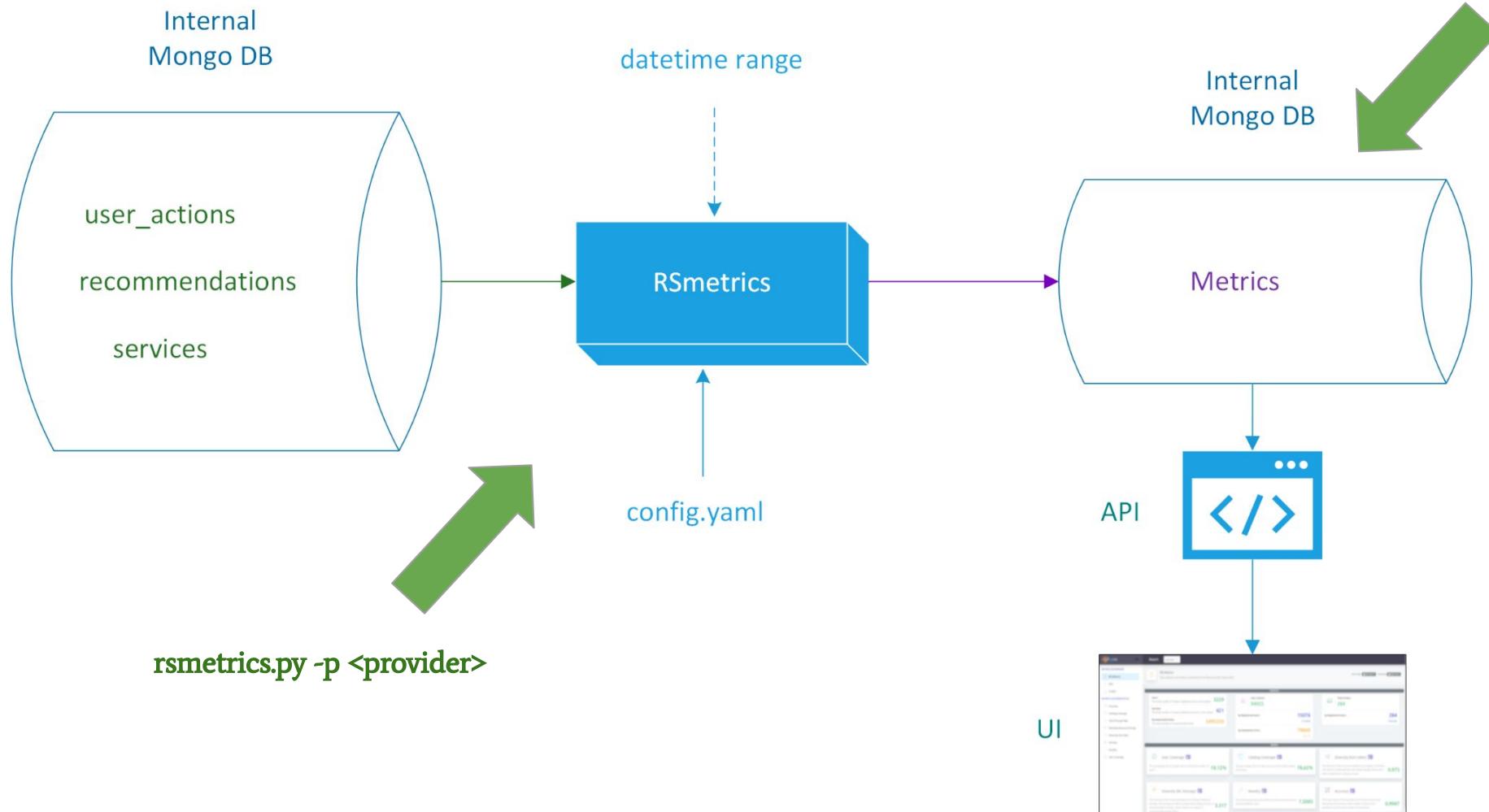
RS Metrics Unit



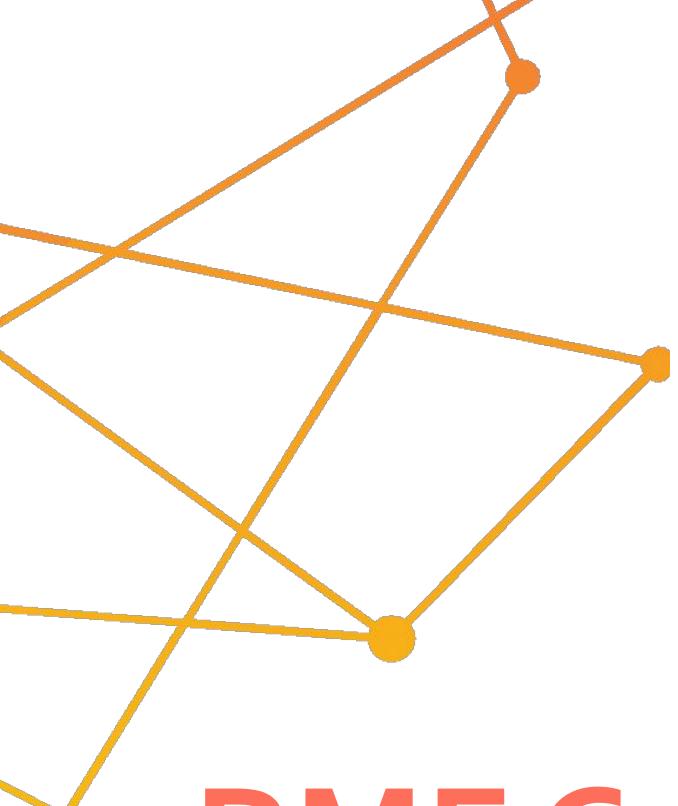
RS Metrics Unit



RS Metrics Unit

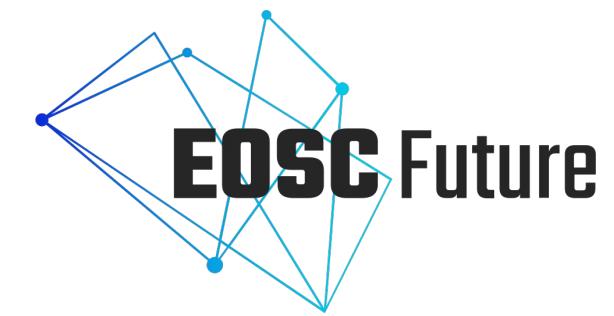


Statistics
Metrics
KPIs
Graphs' Data

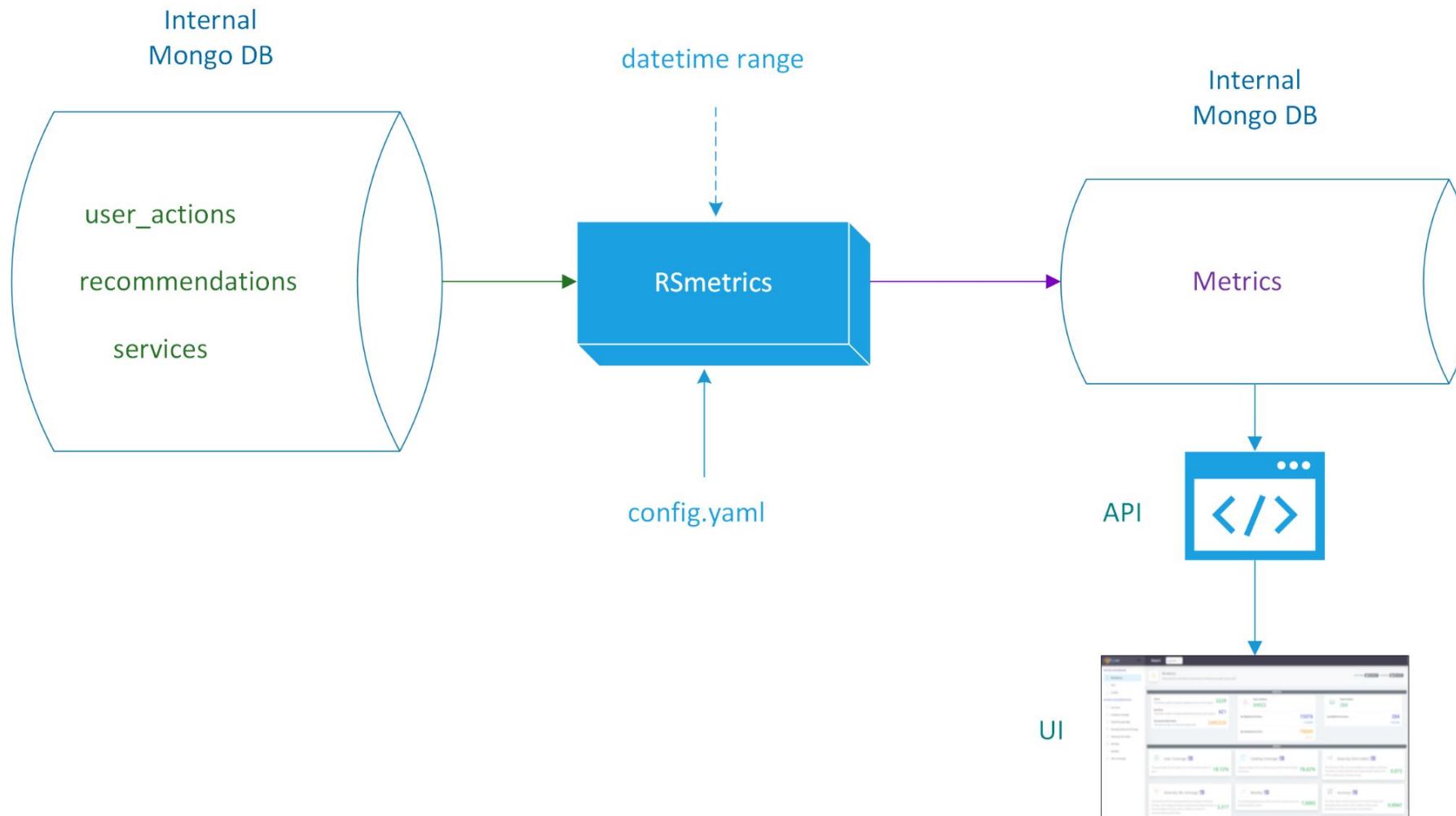


RMF Components

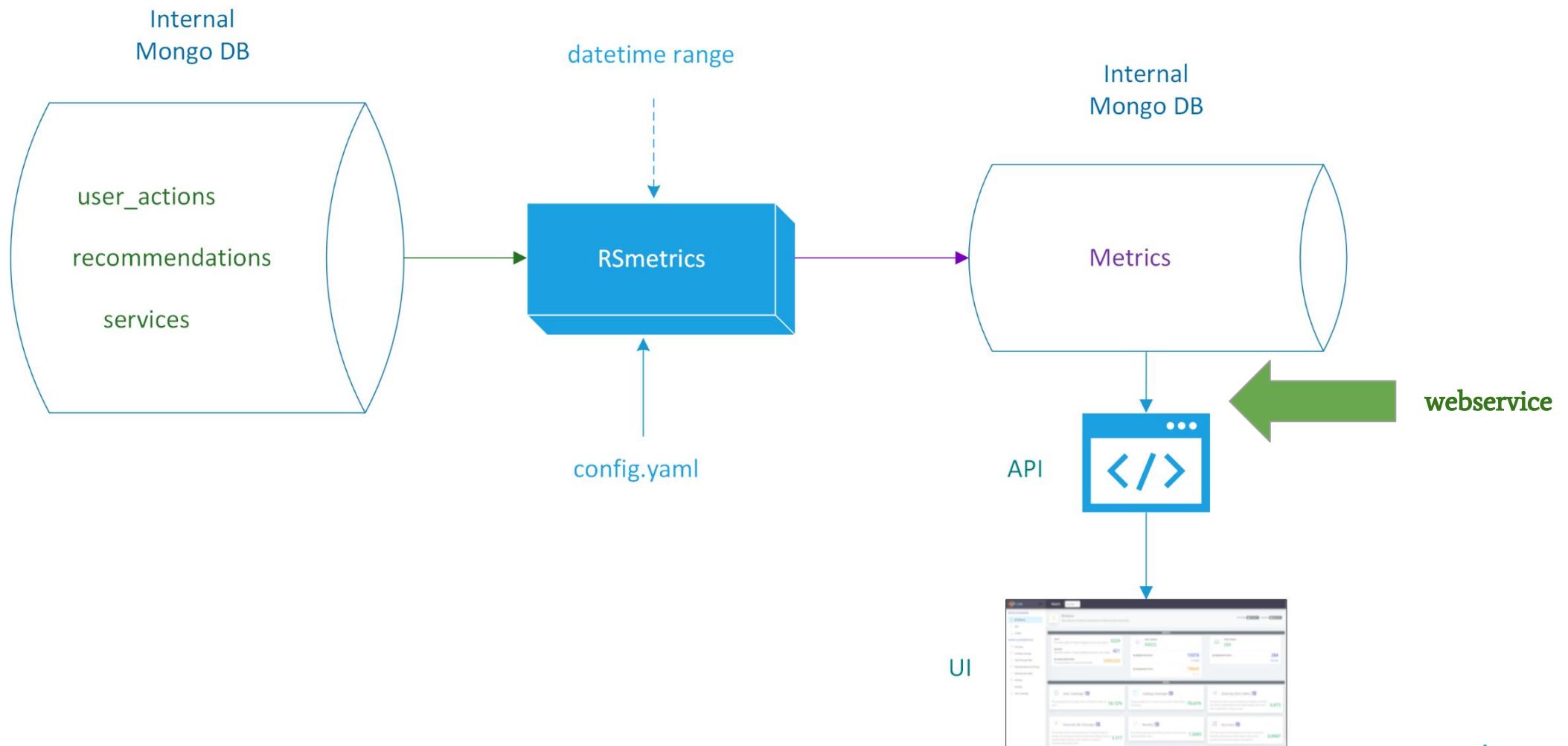
Rest API/ UI Dashboard Unit

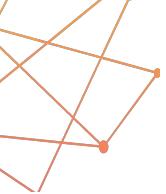


REST API/UI Dashboard Unit

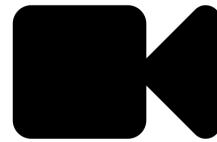


REST API/UI Dashboard Unit

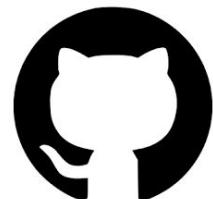




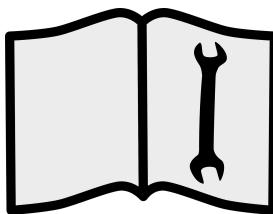
More to see ...



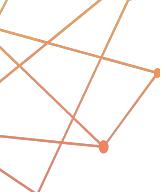
[overview](#)



<https://github.com/ARGOeu/eosc-recommender-metrics>

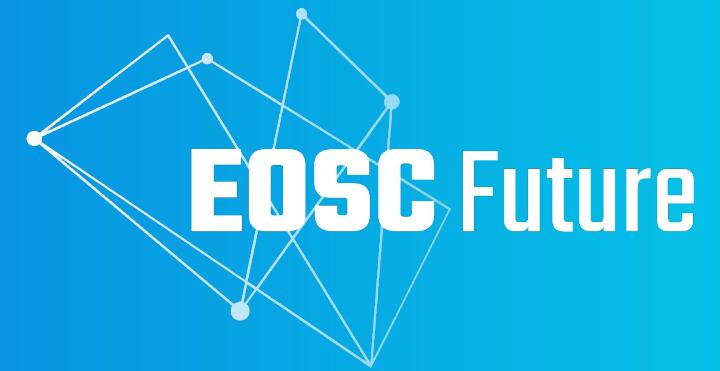


<https://argoeu.github.io/eosc-recommender-metrics/docs>



Ready to answer your questions!





Thank you for your attention

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