

# Recommender Metrics Framework

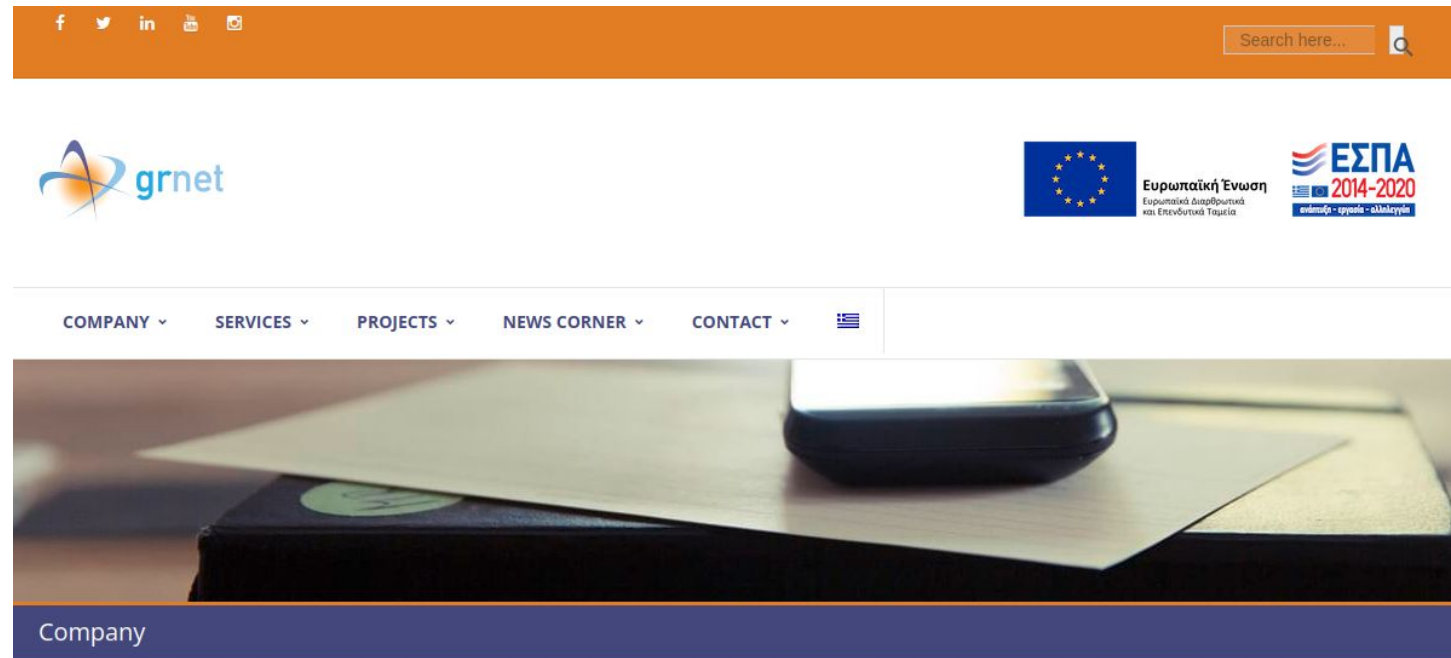
Measuring the success of a Recommender System

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National Infrastructures for Research and Technology (GRNET)

# 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>



# 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 ARG0eu. The organization's profile includes a logo of an eye and the tagline: "A team working with the latest technologies about accounting, monitoring, messaging and eSeal Capabilities". Navigation tabs include Overview, Repositories (92), Projects, Packages, and People (3). The "Pinned" section features four repositories: gr.gnet.eesal (E-signature library, Java, 1 star, 6 forks), argo-messaging (The ARG0 Messaging API is implemented as a Publish/Subscribe Service. Instead of focusing on a single Messaging API specification for handling the logic of publishing/subscribing to the broker netw..., Go, 5 stars, 10 forks), argo-accounting (ARGO Accounting for Services, Java, 2 stars, 6 forks), and argo-monitoring (The main repository to document the argo monitoring, JavaScript, 4 forks). The "Repositories" section has a search bar and filters for Type, Language, and Sort. It lists three repositories with commit history graphs: argo-api-authn (Authentication Server for ARG0 apis, Go, 0 stars, 5 forks, 0 commits, 7 watchers, updated 8 hours ago), argo-scg (Python, 0 stars, 1 fork, 0 commits, 0 watchers, updated yesterday), and eossc-recommender-metrics (A framework for counting the recommender metrics).



# 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.


Because you watched shows about Anti-Heroes and Moral Ambiguity >




Because you watched shows with Sharp Humor and Strong Female Leads >



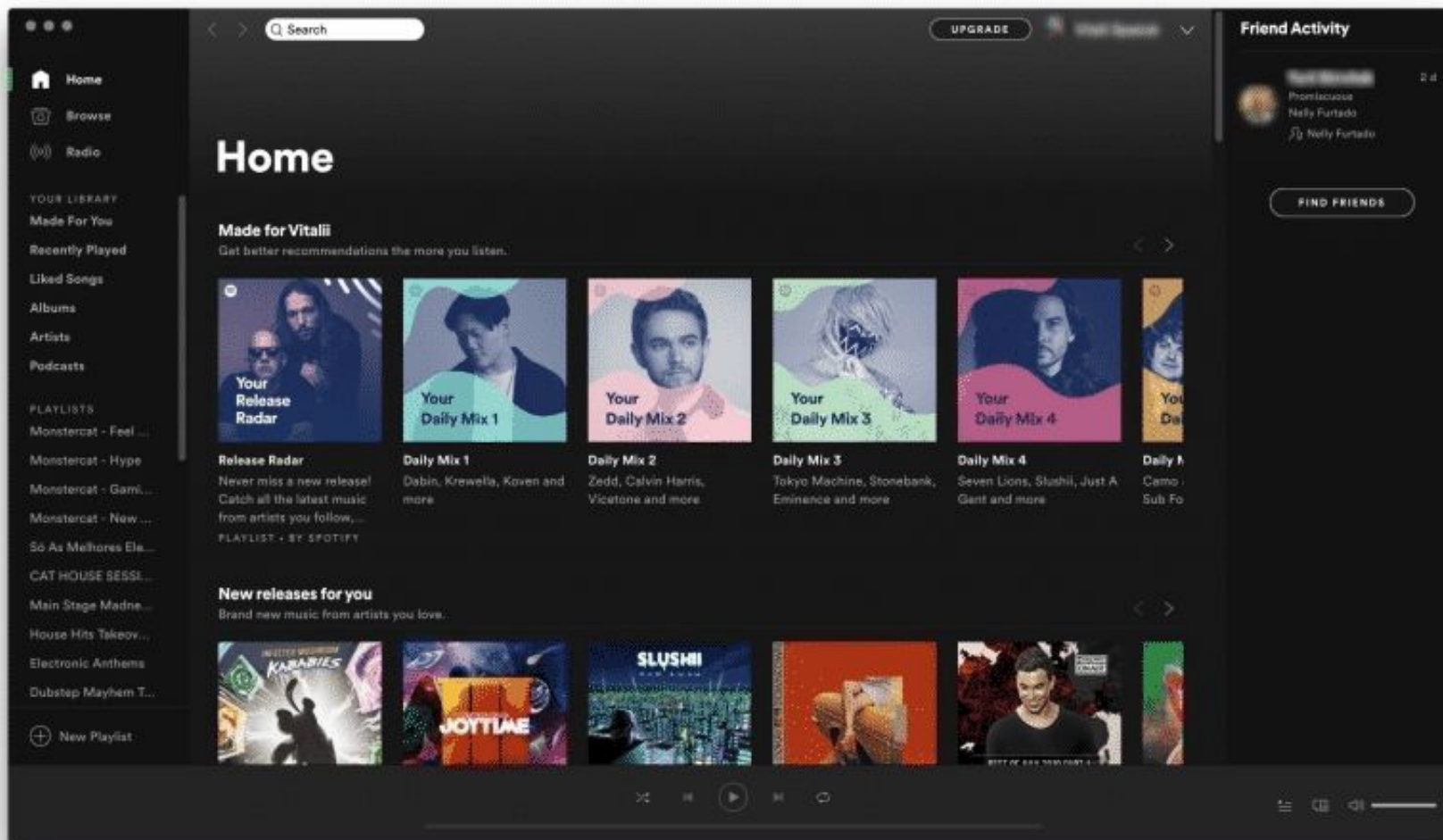
Because you watched shows about Dangerous Worlds and Complex Consequences >



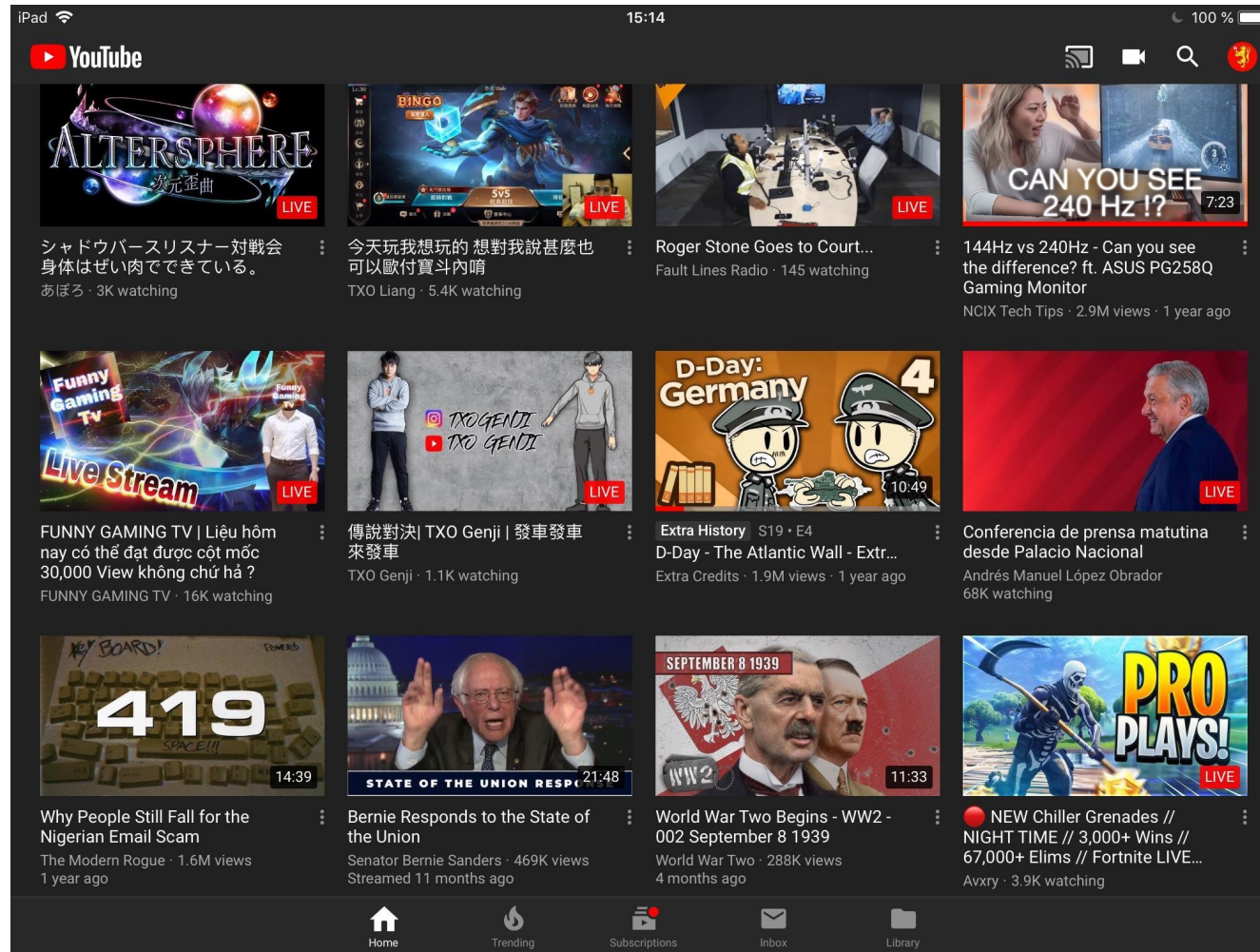
Because you watched shows about Edgy Coming of Age Tales >



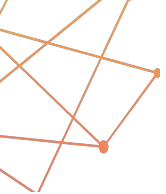
More...



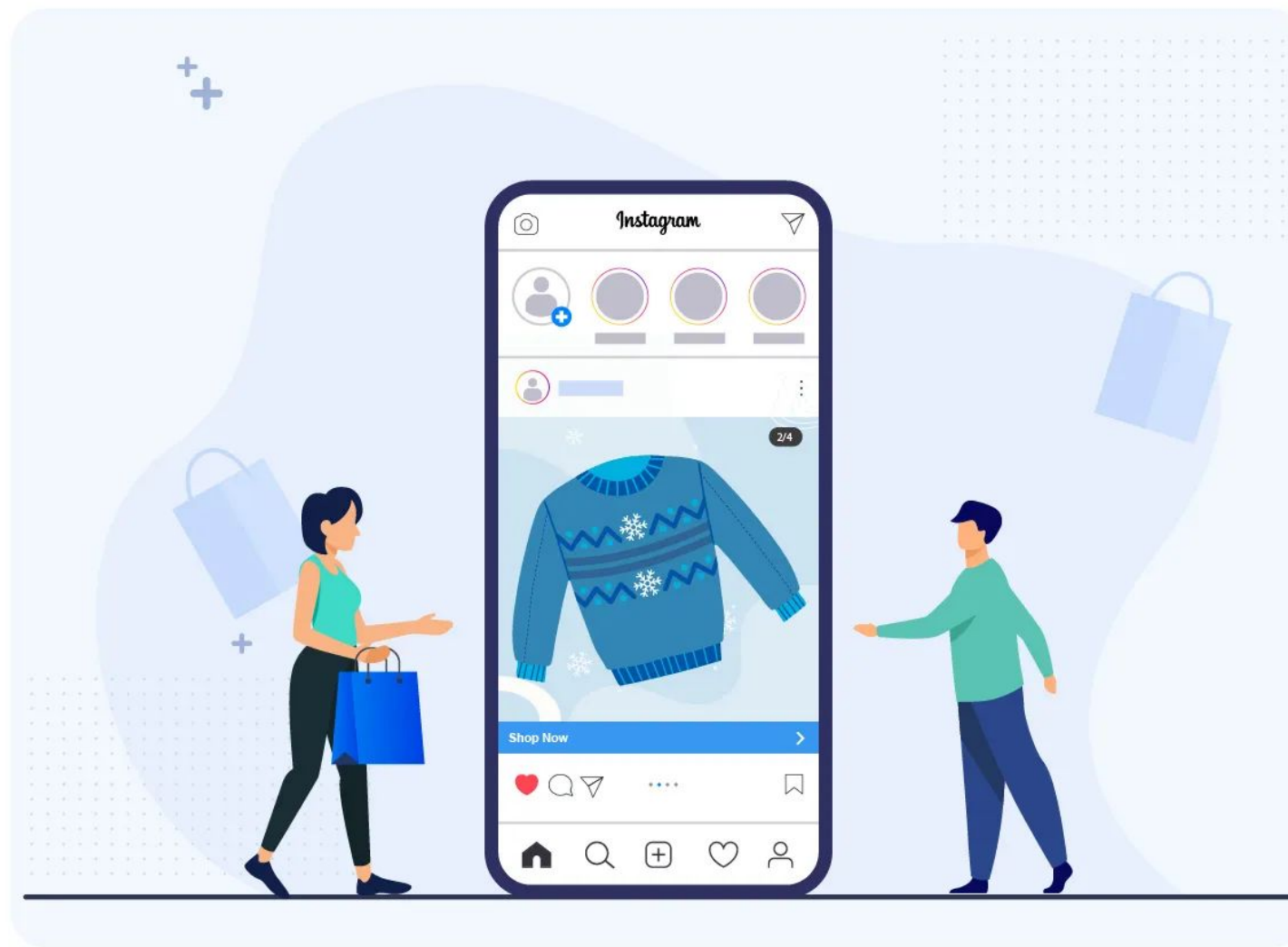
# More...







# More...





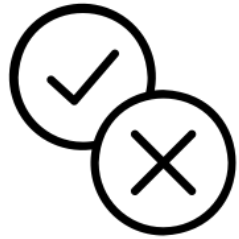
**But...**



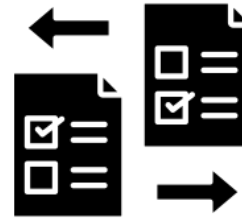
**. How “good” recommendations are?**

**. 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

Since 2023

Since 2022

Since 2019

Custom range...

Sort by relevance

Sort by date

Any type

Review articles

include patents

include citations

Create alert

[HTML] How good your **recommender system** is? A survey on evaluations in **recommendation**

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 ...

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[HTML] springer.com

[HTML] SemCiR: A citation **recommendation system** based on a novel semantic distance **measure**

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] emerald.com

Evaluating **recommendation systems**

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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[PDF] psu.edu

Social network based **recommendation systems**: A short survey

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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[PDF] ieee.org

Related searches

**similarity** measure recommendation system

**systematic review** recommender systems

**collaborative filtering** recommender systems

**social network based** recommender systems

recommendation systems **for software engineering**

**business opportunities** recommendation systems

**good** recommendations recommender systems

**accuracy metrics** recommender systems

Dimensions and metrics for evaluating **recommendation systems**

[PDF] github.io

# How to measure success?



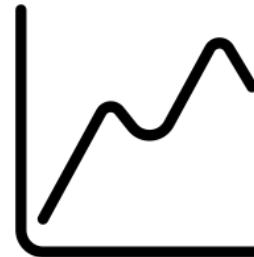
**Statistics:** quantifies the occurrences of various data entities, such as user interactions, item popularity, or recommendation relevance scores.



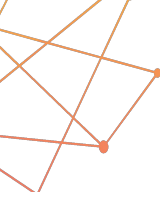
**Metrics:** goes beyond simple **counts** and offer sophisticated characterizations of Recommender System's performance.



**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.



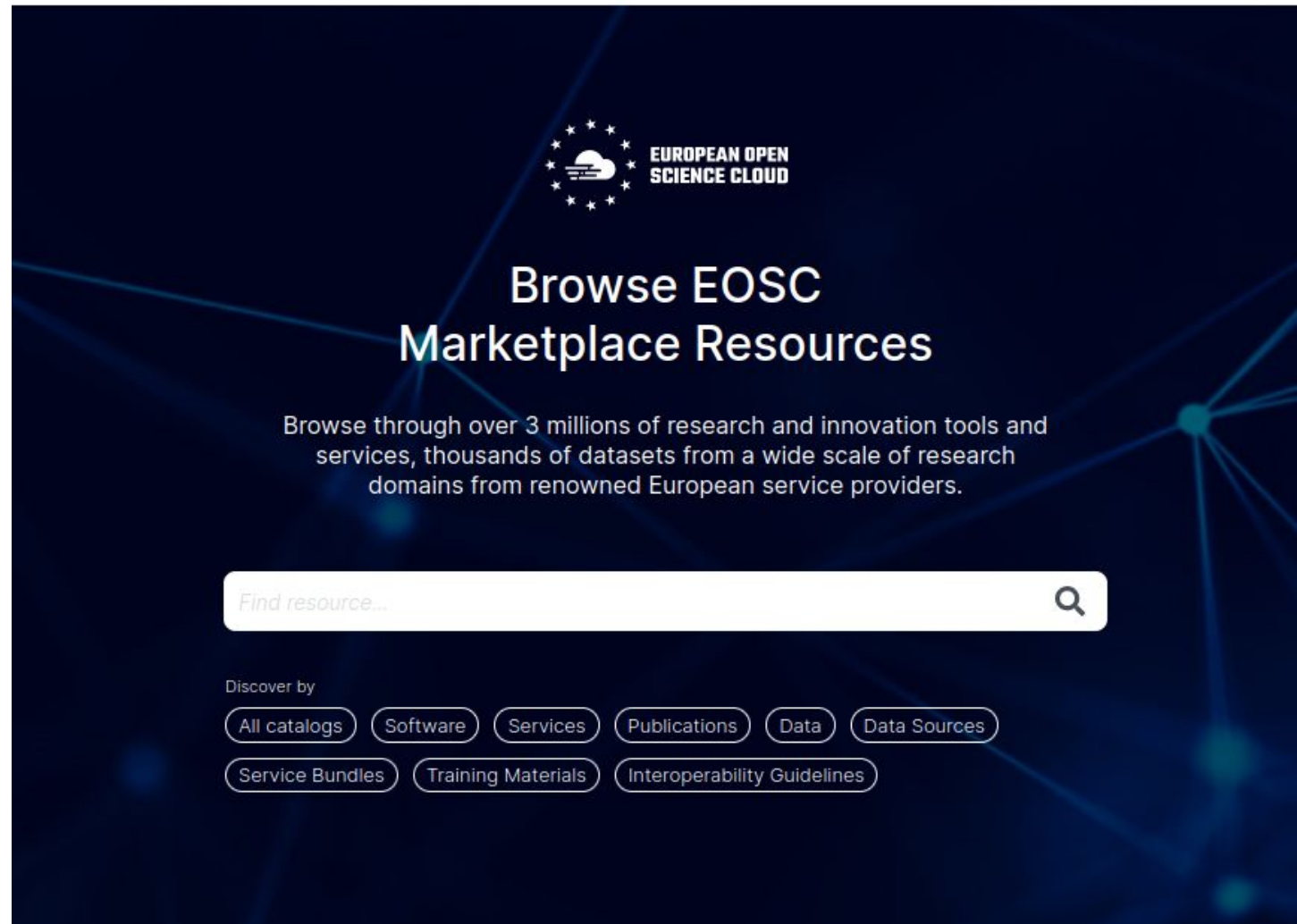
# What is EOSC?

- 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.



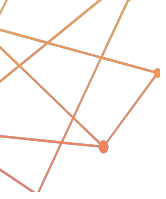
The screenshot shows the homepage of the EOSC Marketplace. At the top, there is a navigation menu with links: "About EOSC", "Browse Marketplace" (which is underlined), "Providers Hub", "Monitoring", "Status", and "Contact us". Below the navigation is a dark blue header area with the "EUROPEAN OPEN SCIENCE CLOUD" logo on the left. The main heading reads "Browse EOSC Marketplace Resources". Below this, a text block states: "Browse through over 3 millions of research and innovation tools and services, thousands of datasets from a wide scale of research domains from renowned European service providers." A search bar with the placeholder text "Find resource..." and a magnifying glass icon is positioned below the text. Underneath the search bar, there is a "Discover by" section with several filter buttons: "All catalogs", "Software", "Services", "Publications", "Data", "Data Sources", "Service Bundles", "Training Materials", and "Interoperability Guidelines".



# 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

## 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.

CAN BE USED

## 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



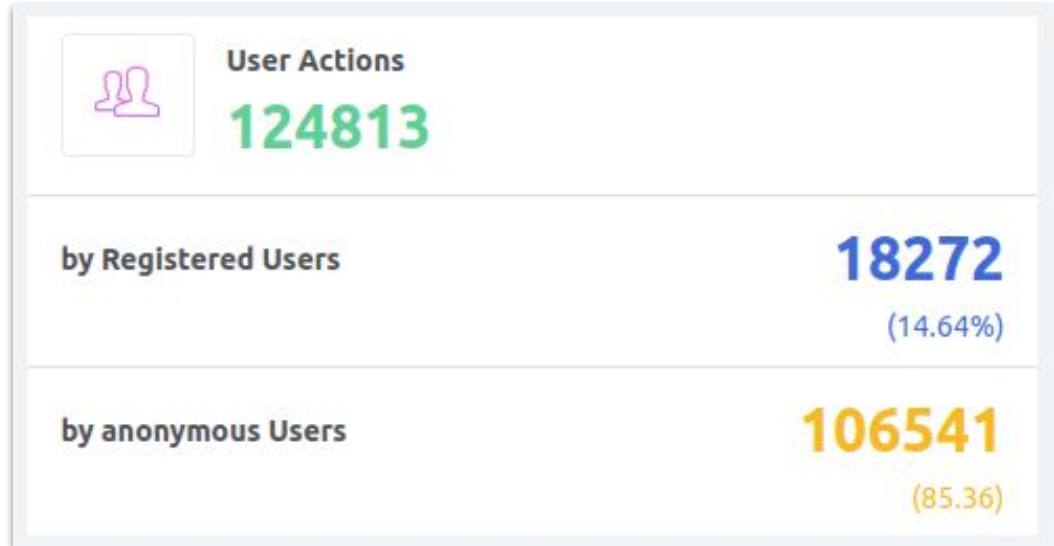


# What it offers?

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

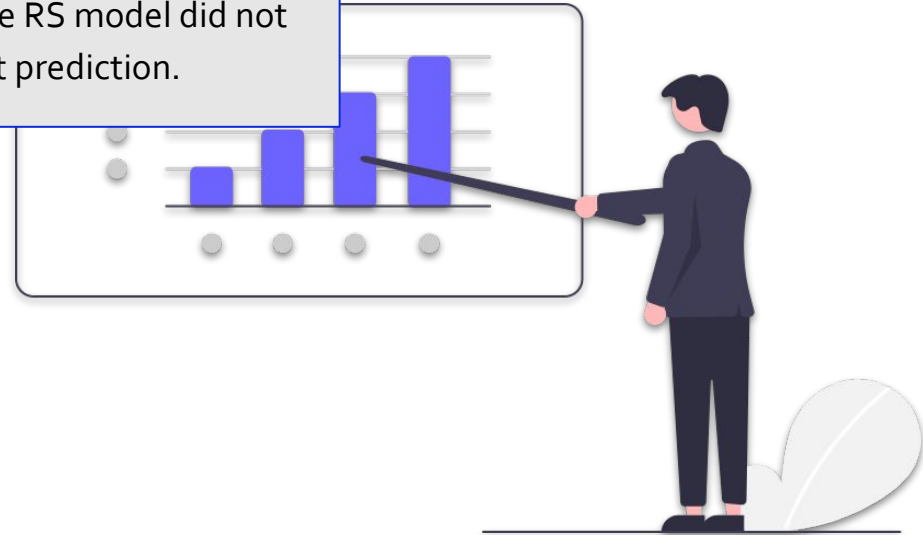


# What it offers?

## Metrics

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

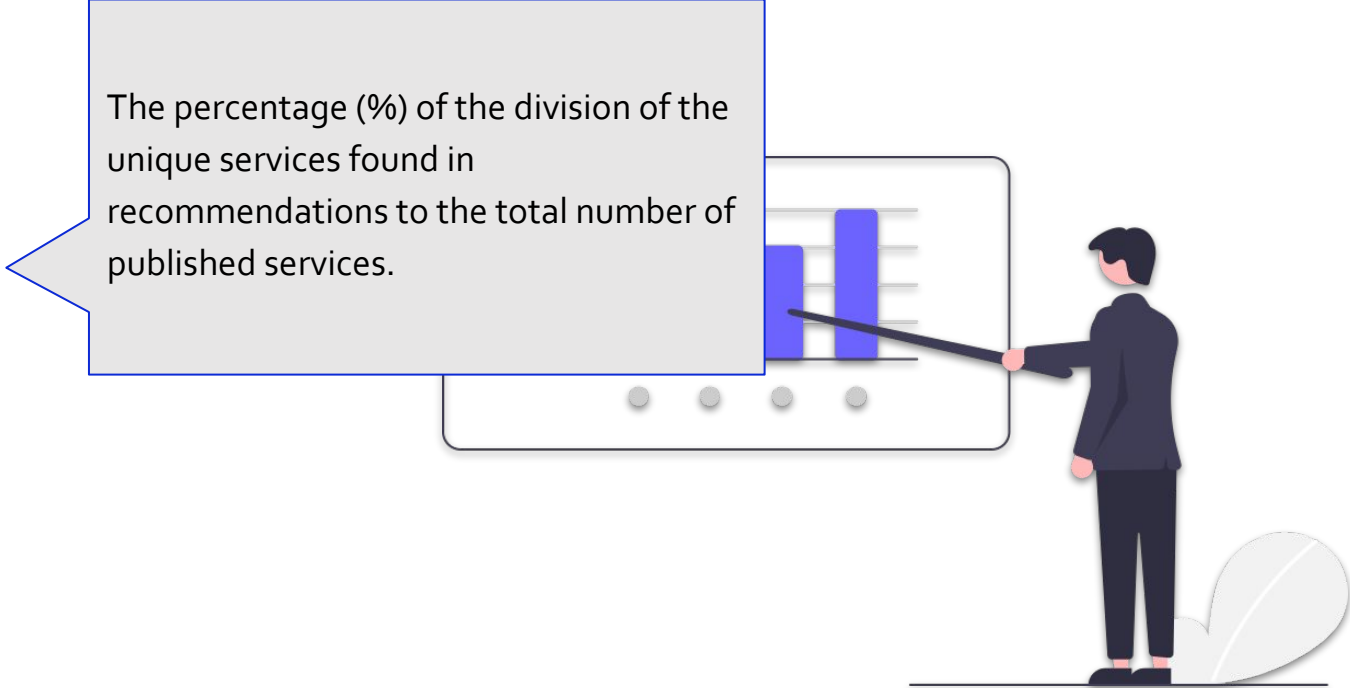




# What it offers?

## Metrics

- Accuracy
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- Diversity Gini Index
- Diversity Shannon Entropy
- Novelty
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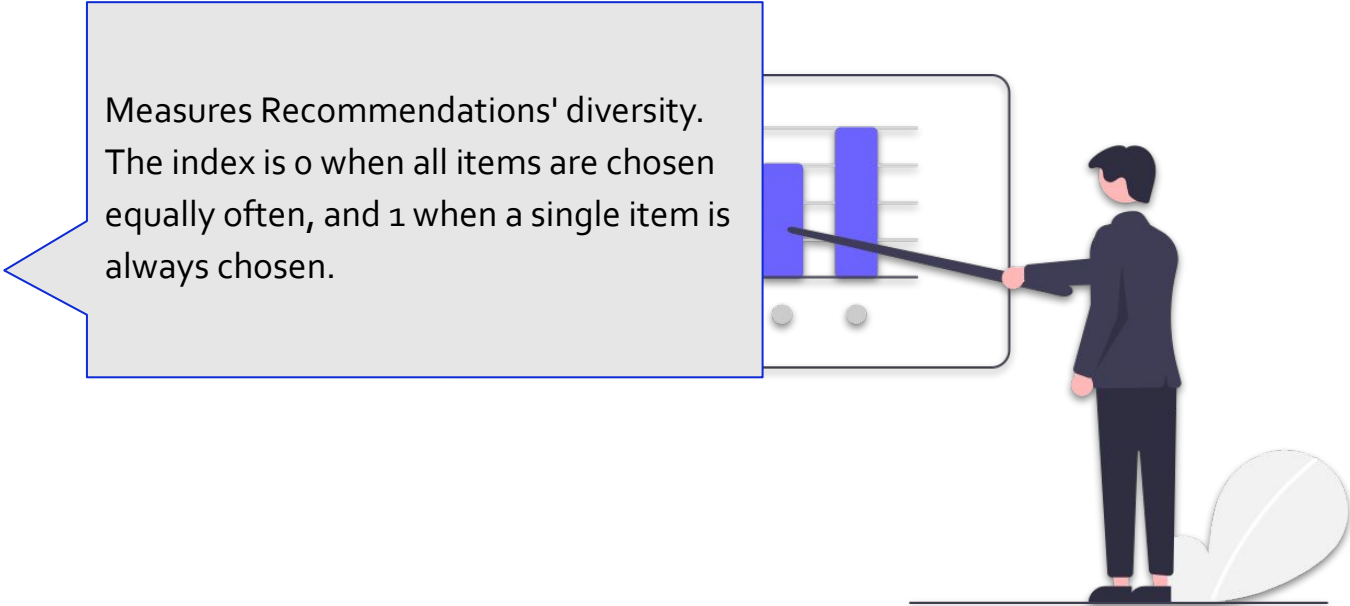
The percentage (%) of the division of the unique services found in recommendations to the total number of published services.



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

Measures Recommendations' diversity. The entropy is 0 when a single item is always chosen or recommended, and  $\log n$  when  $n$  items are chosen or recommended equally often.



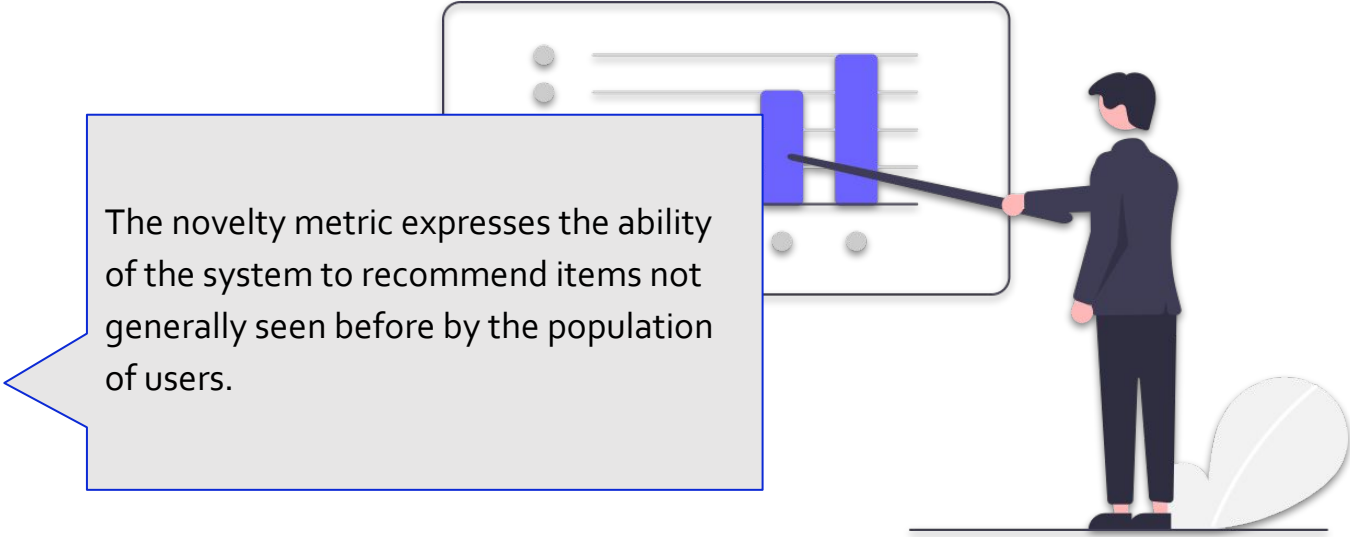




# What it offers?

## Metrics

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



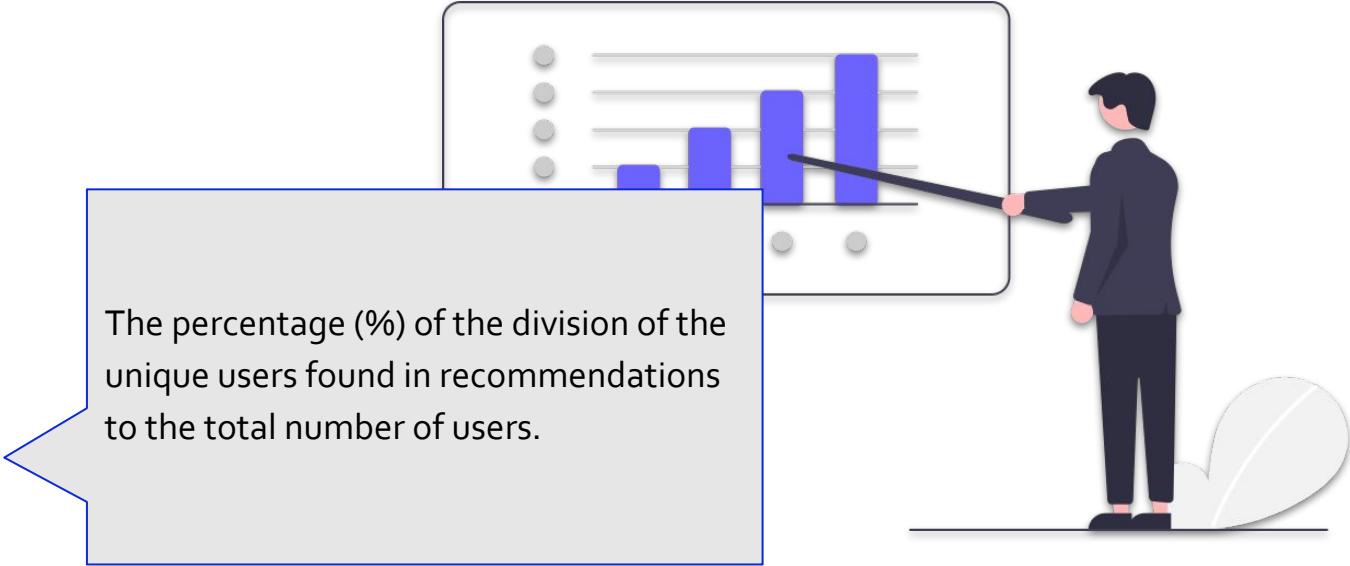
The novelty metric expresses the ability of the system to recommend items not generally seen before by the population of users.



# What it offers?

## Metrics

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



The percentage (%) of the division of the unique users found in recommendations to the total number of 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

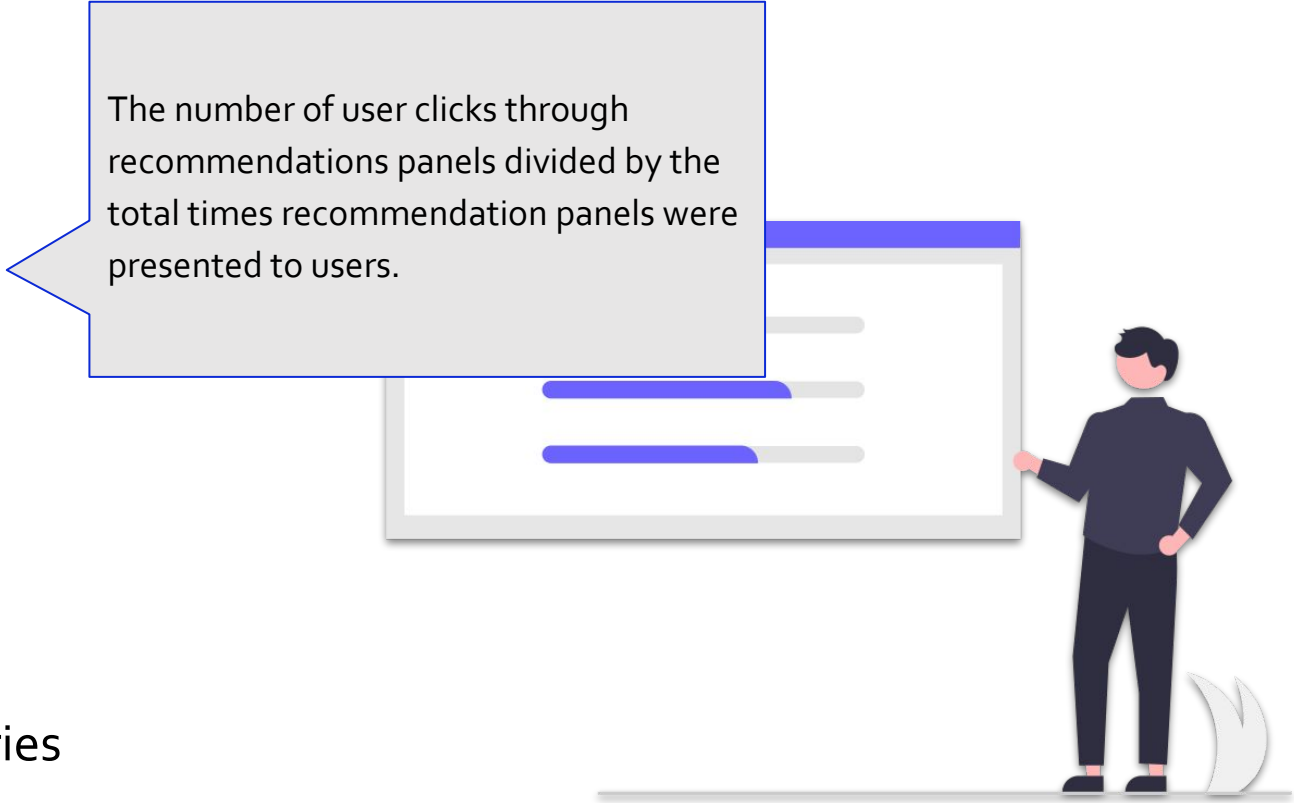




# What it offers?

## 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



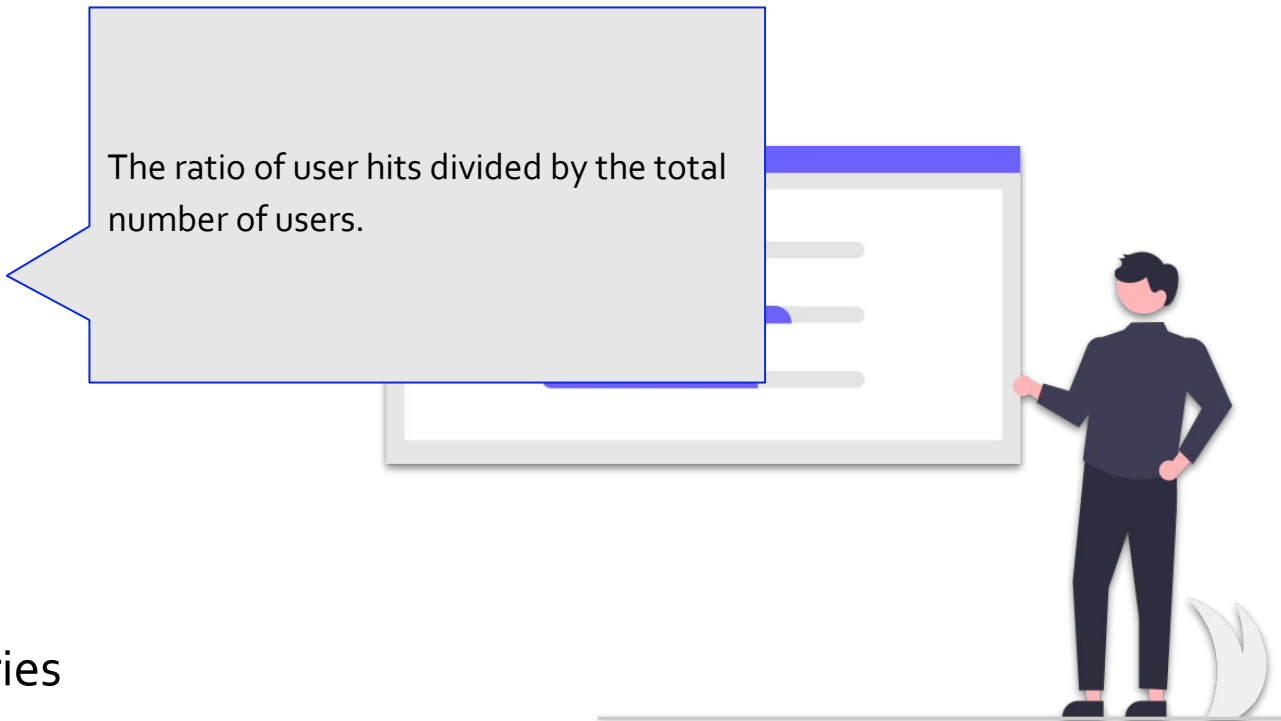
The number of user clicks through recommendations panels divided by the total times recommendation panels were presented to users.



# 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



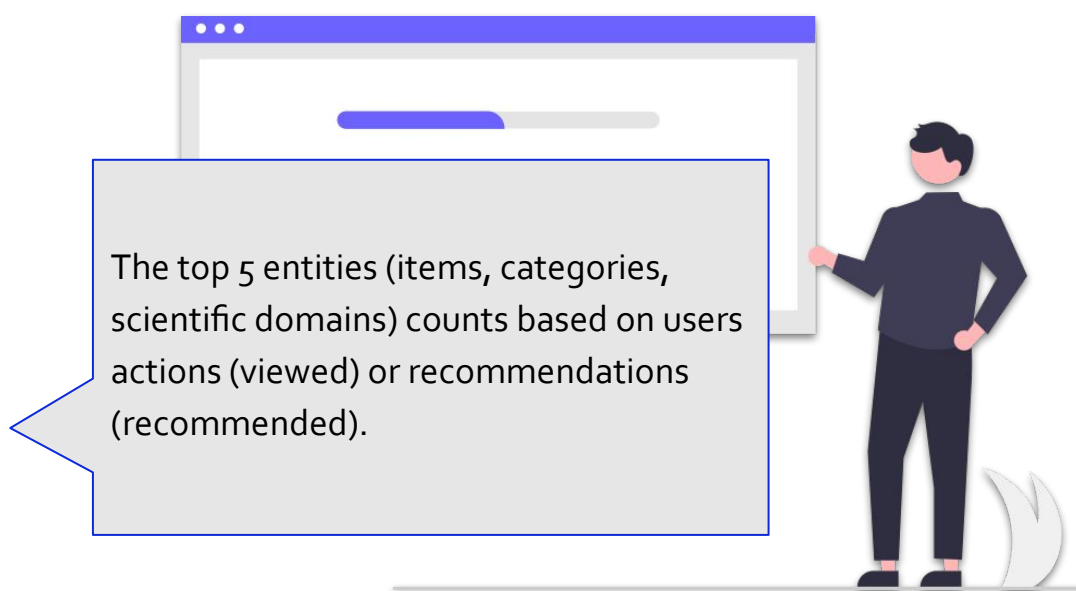
The ratio of user hits divided by the total number of users.



# 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**



The top 5 entities (items, categories, scientific domains) counts based on users actions (viewed) or recommendations (recommended).

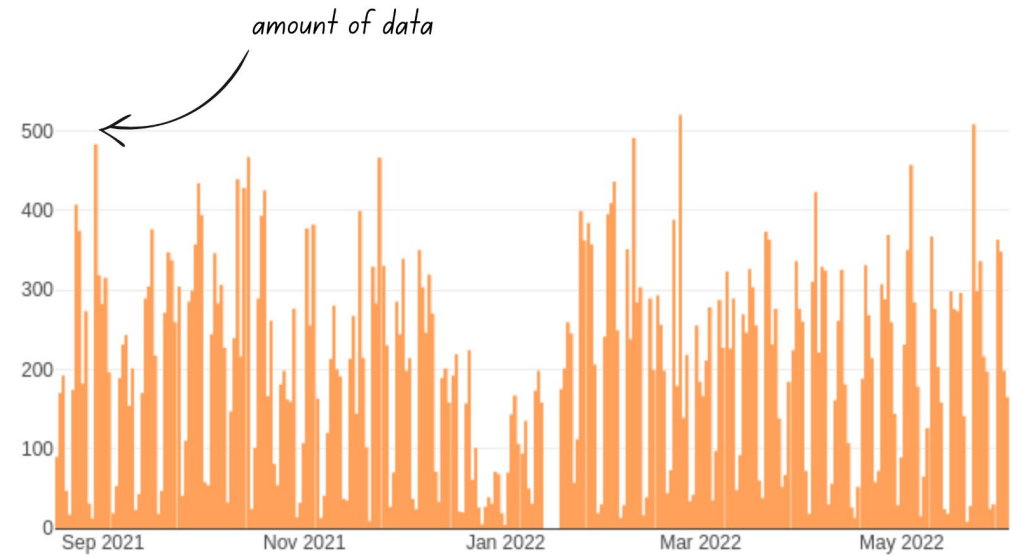
# What it offers?

## Graphs

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

USER ACTIONS PER DAY

*type of metric*



*referenced date-time range*



# What it offers?

## Rest API

- Statistics
- Metrics
- KPIs
- Graphs' Data

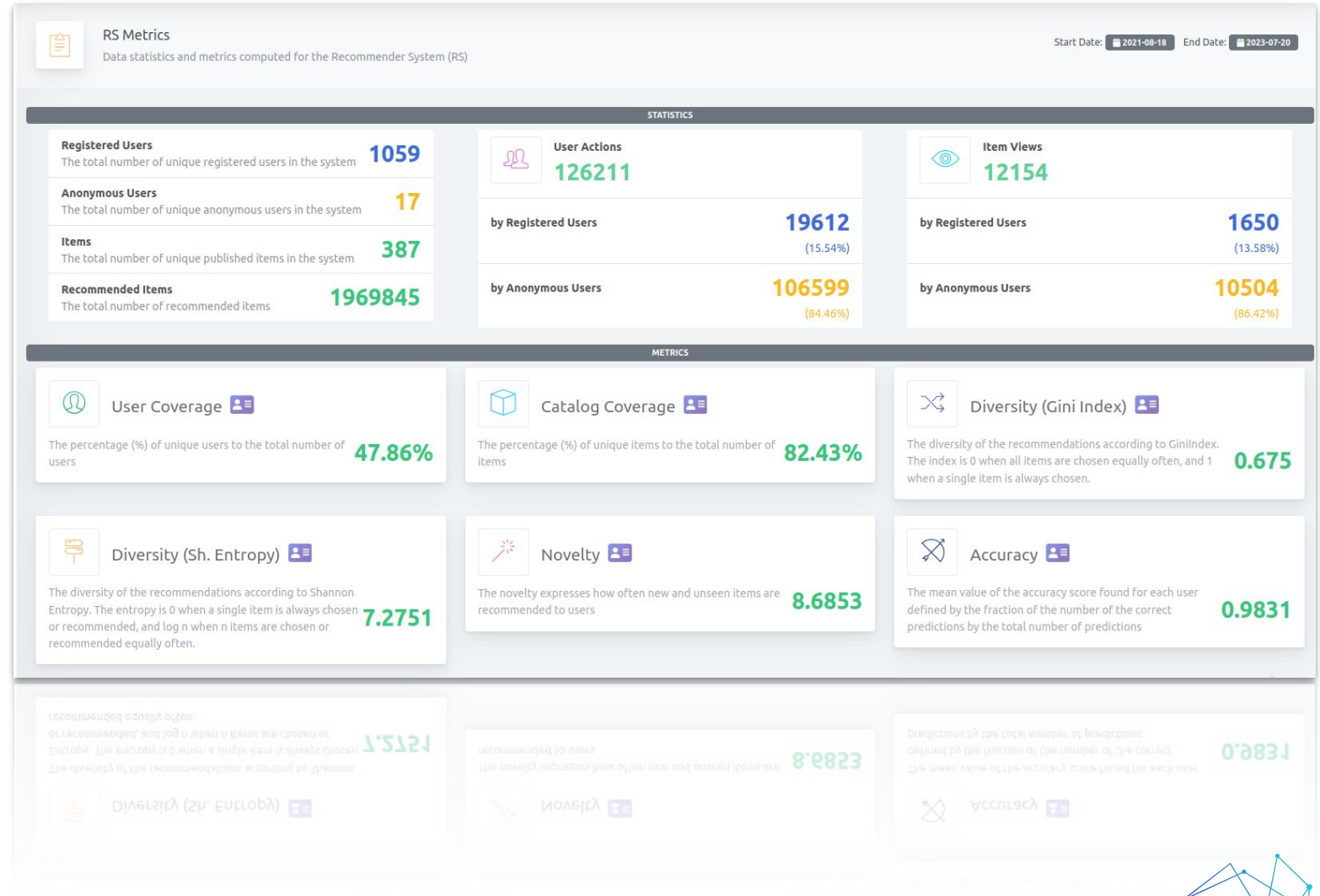
```
▼ 0:
  name: "accuracy"
  value: 0.9942
  doc: "The mean value of the accuracy score found for each user defined by the fraction of
▼ 1:
  name: "catalog_coverage"
  value: 82.43
  doc: "The percentage (%) of unique services to the total number of services"
▼ 2:
  name: "click_through_rate"
  value: 0.03
  doc: "The number of user clicks through recommendations panels divided by the total times
▼ 3:
  name: "diversity"
  value: 3.2668
  doc: "The diversity of the recommendations according to Shannon Entropy. The entropy is 0
▼ 4:
  name: "diversity_gini"
  value: 0.9718
  doc: "The diversity of the recommendations according to GiniIndex. The index is 0 when al
▼ 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
▼ 6:
  name: "novelty"
  value: 8.0964
  doc: "The novelty expresses how often new and unseen items are recommended to users"
▼ 7:
```



# What it offers?

## UI Dashboard

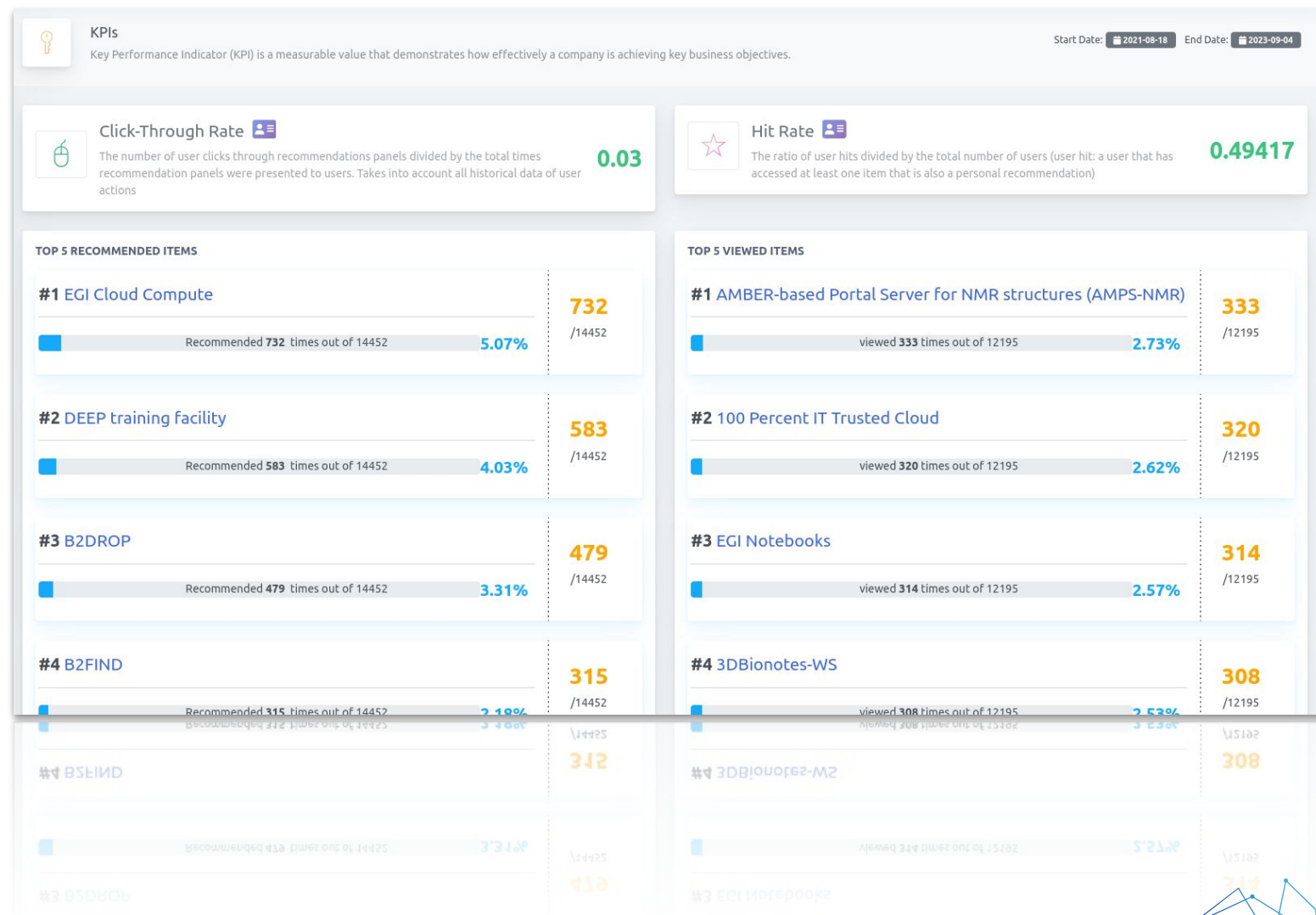
- Statistics
- Metrics
- KPIs
- Graphs



# What it offers?

## UI Dashboard

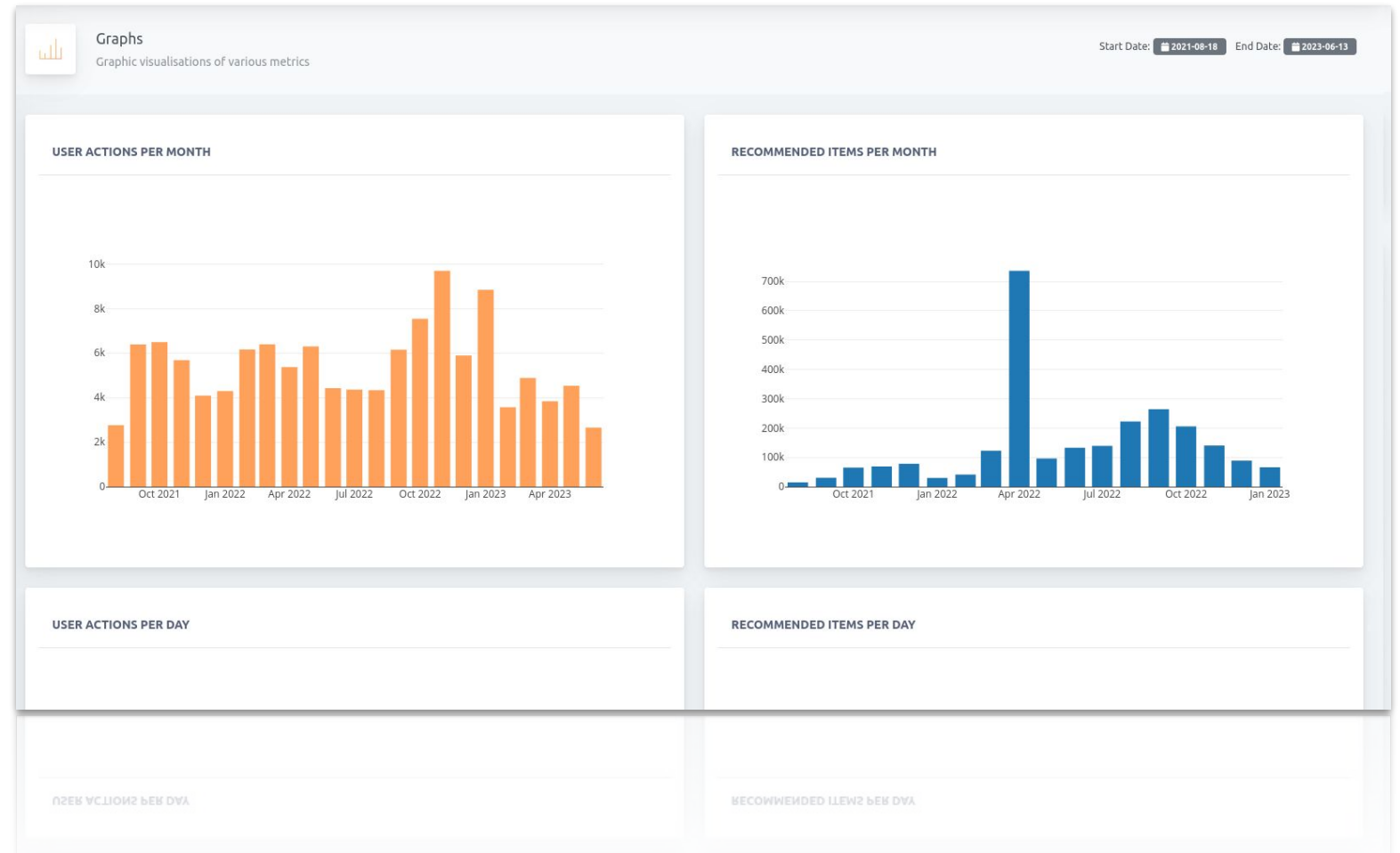
- Statistics
- Metrics
- KPIs
- Graphs



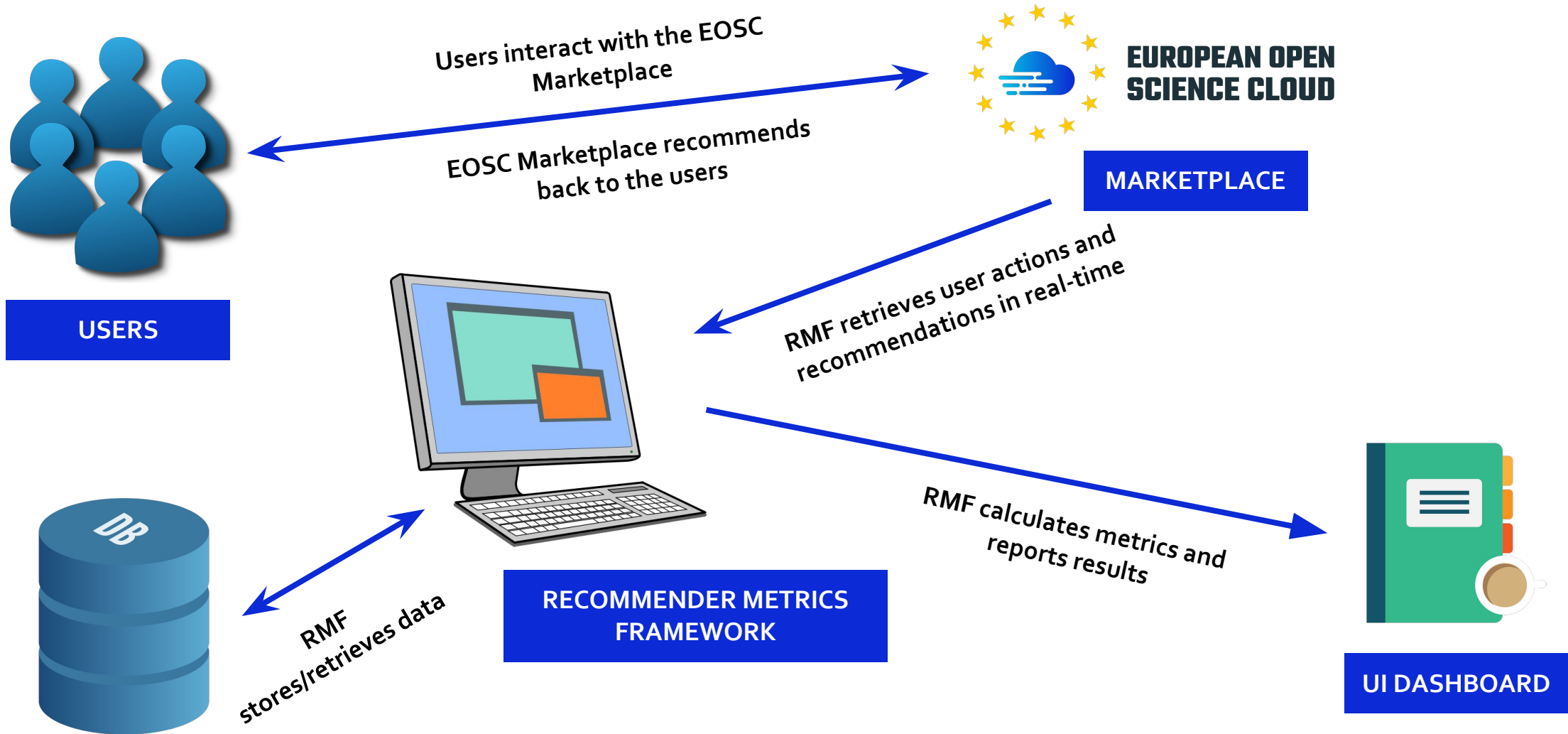
# What it offers?

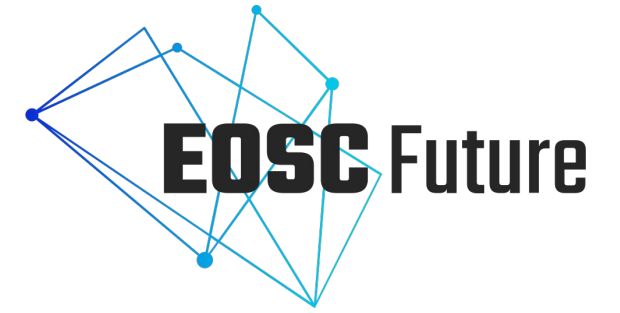
## UI Dashboard

- Statistics
- Metrics
- KPIs
- Graphs



# Process Flow



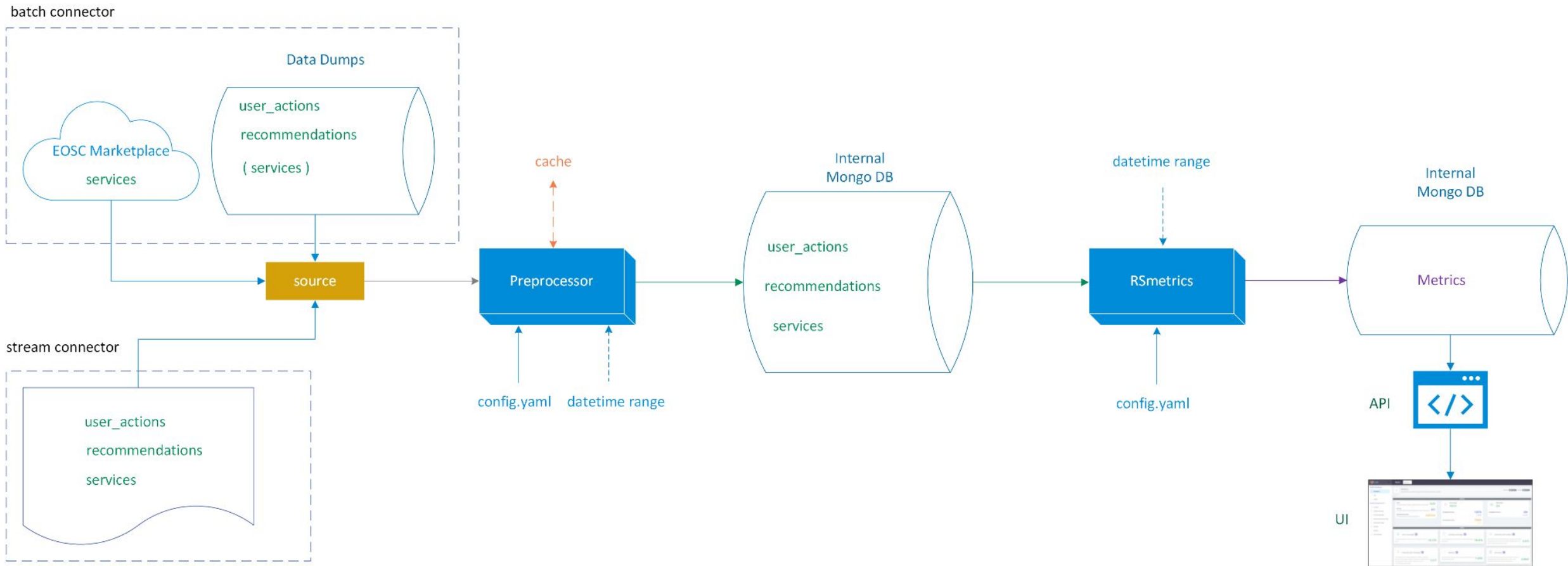


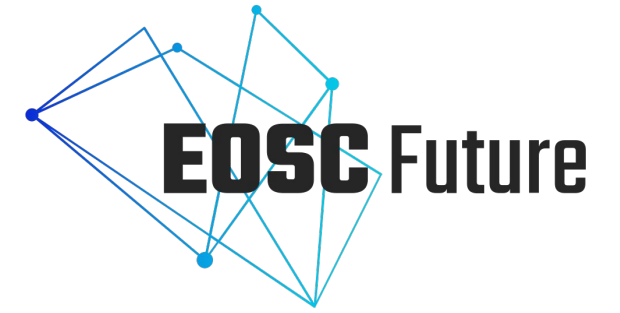
# RMF Components

All Units



# Framework's components

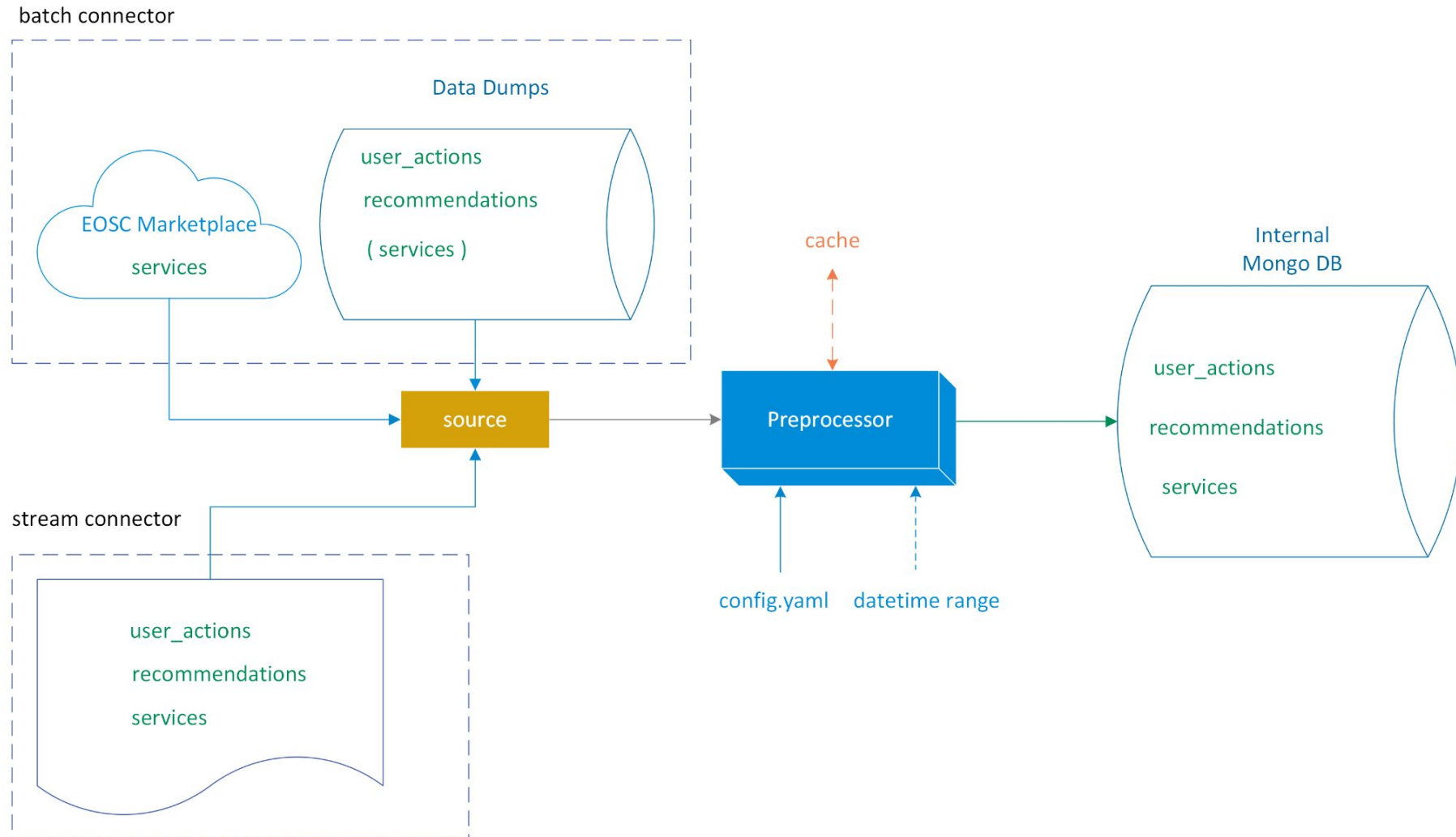




# RMF Components

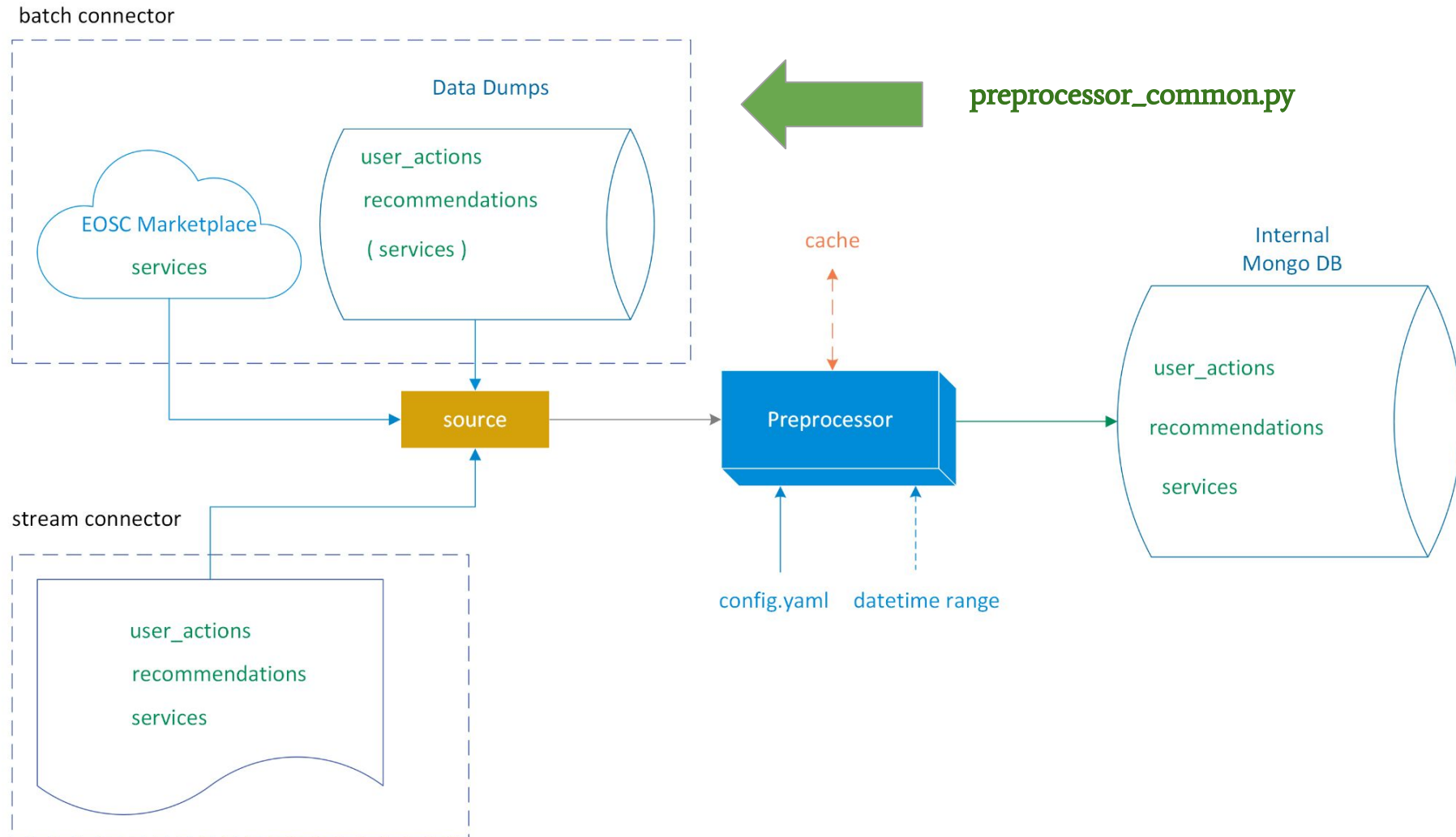
Preprocessor Unit

# Preprocessor Unit

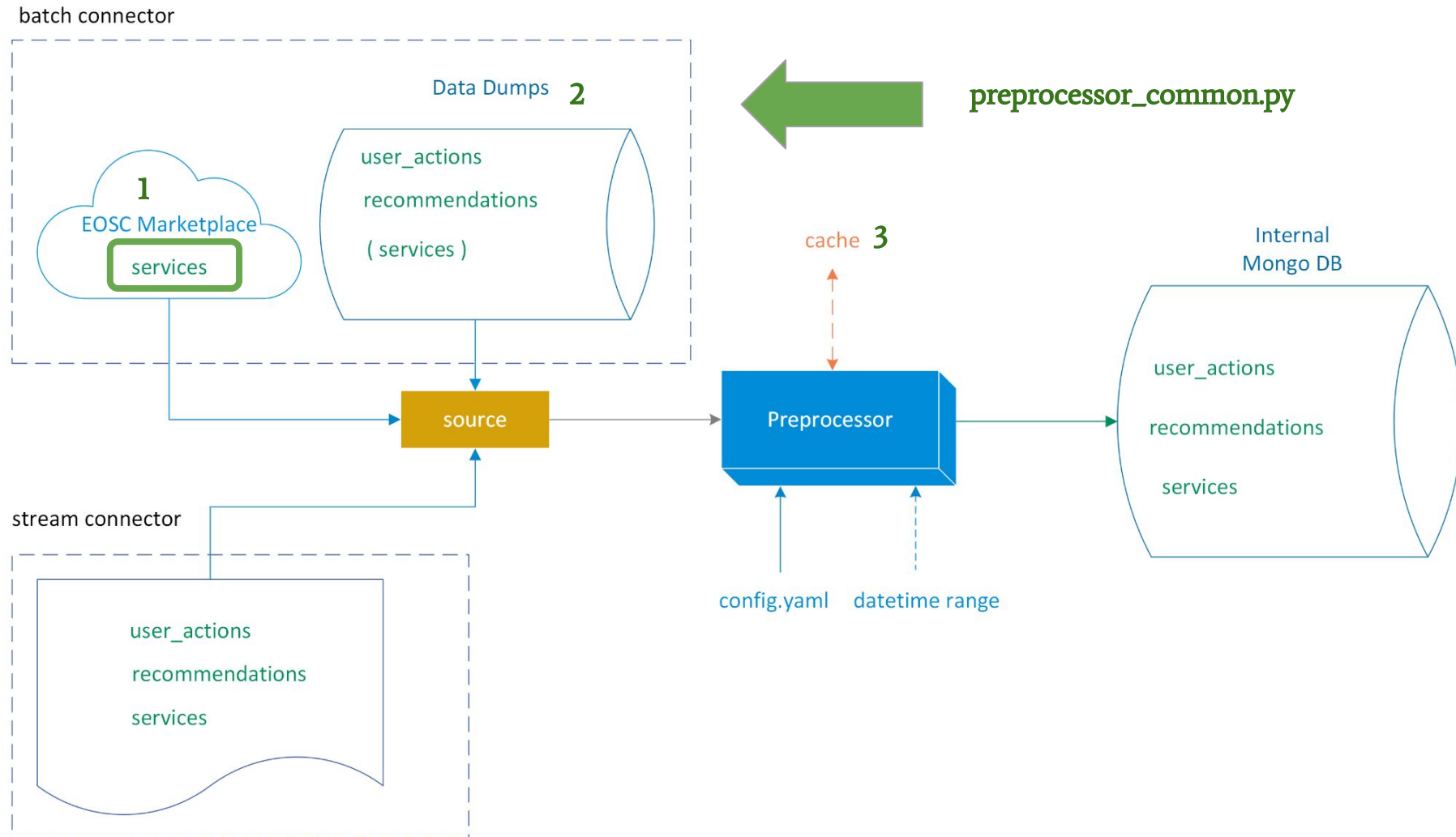




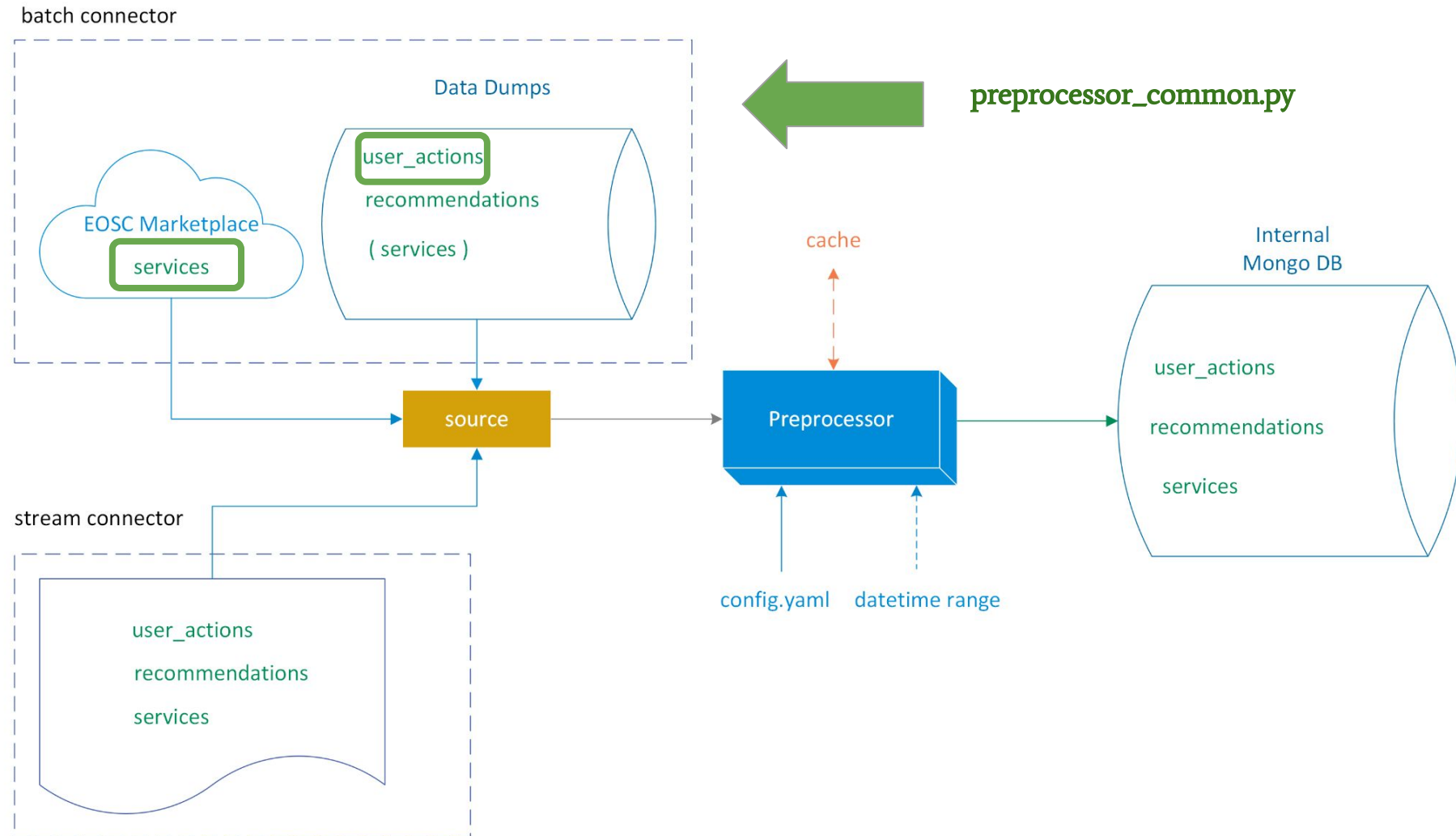
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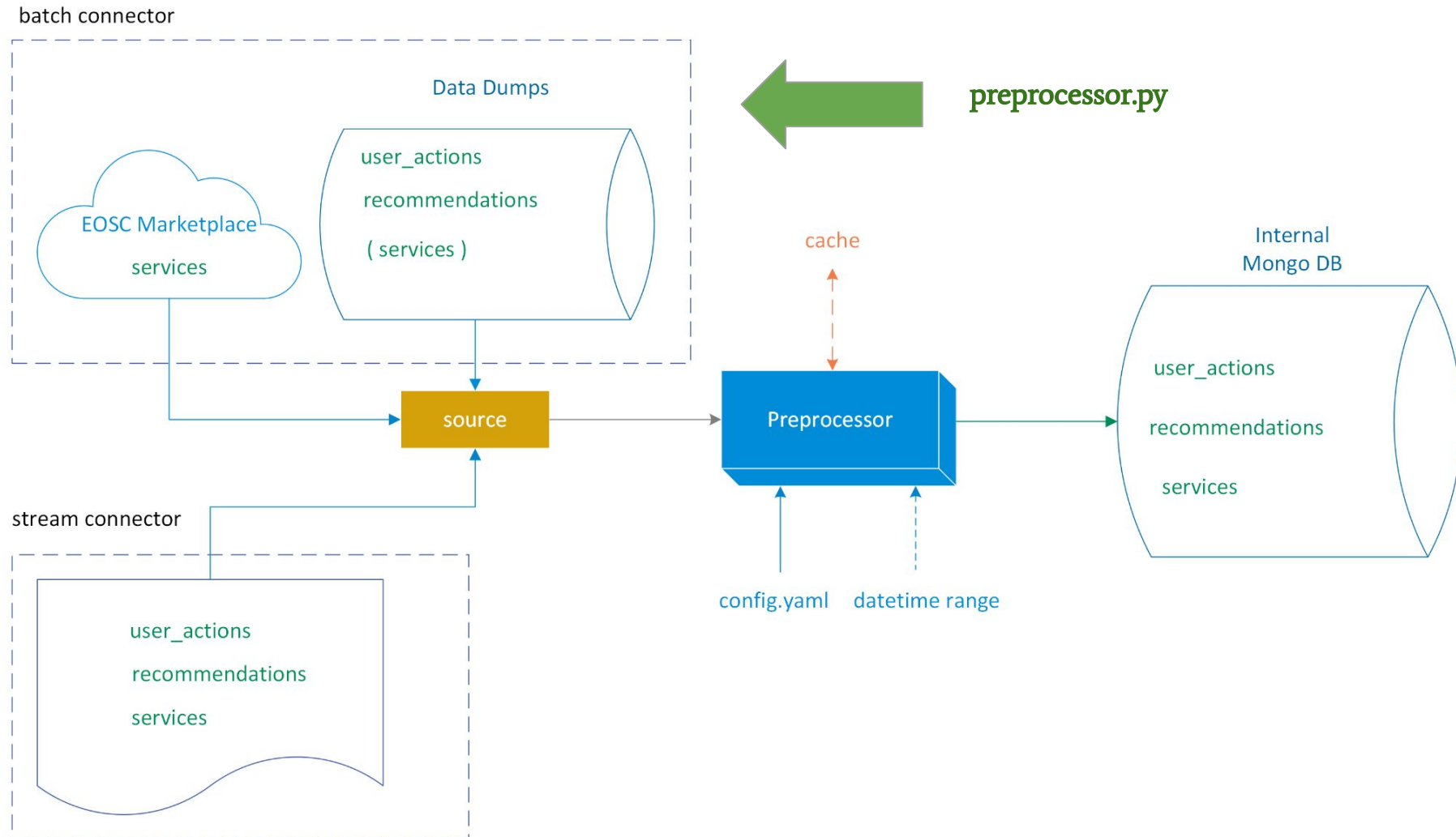
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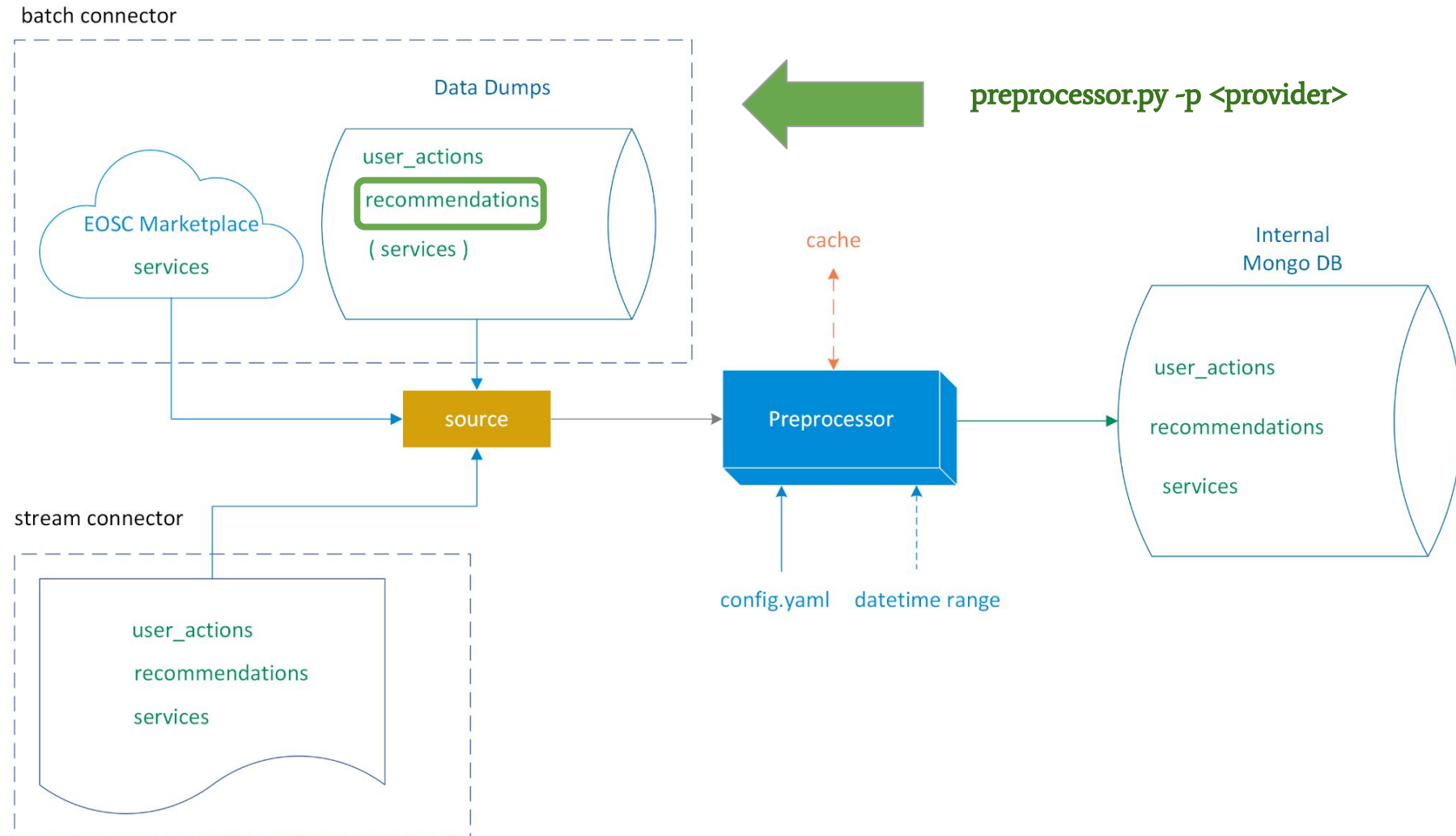
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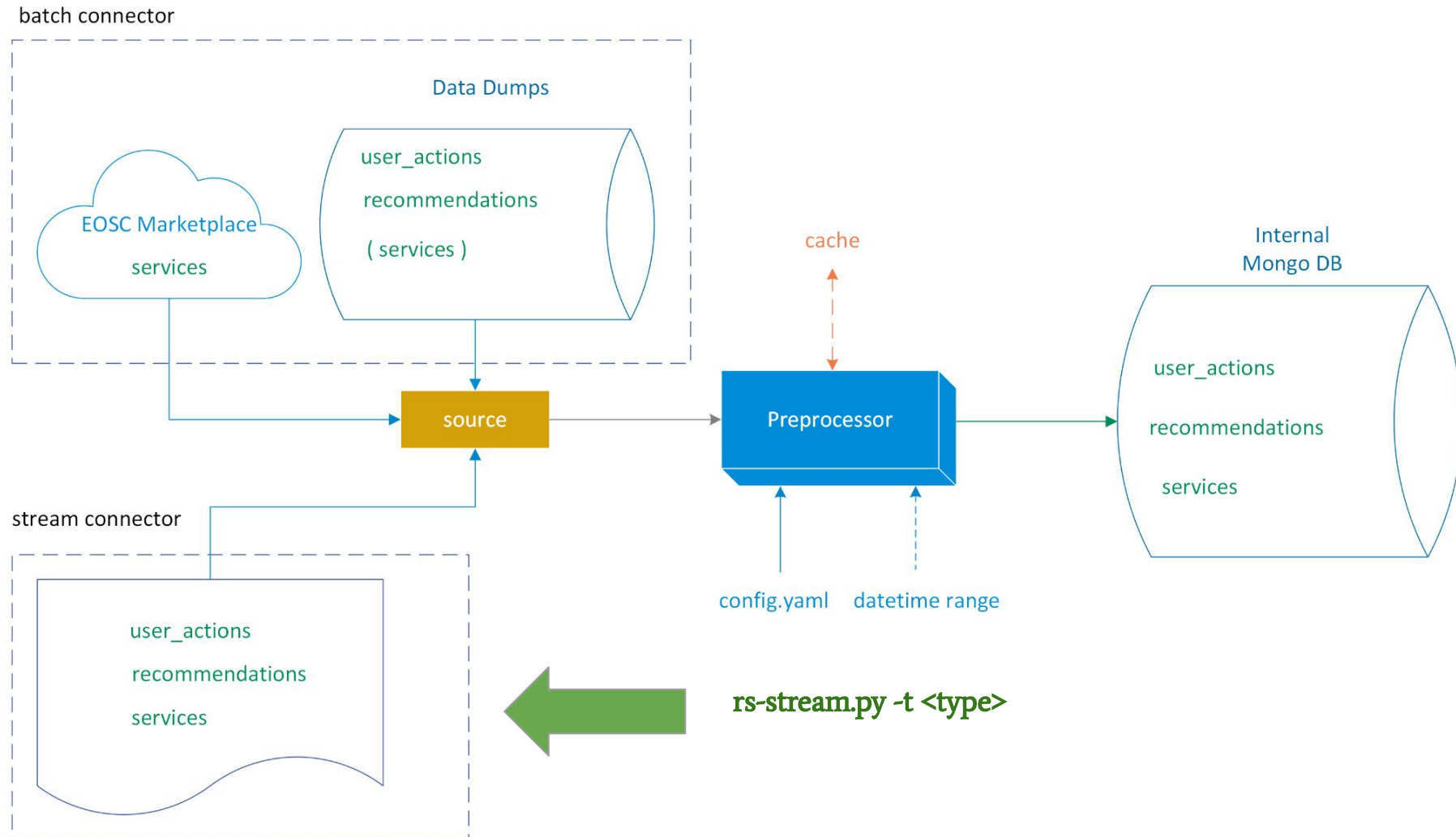
# Preprocessor Unit

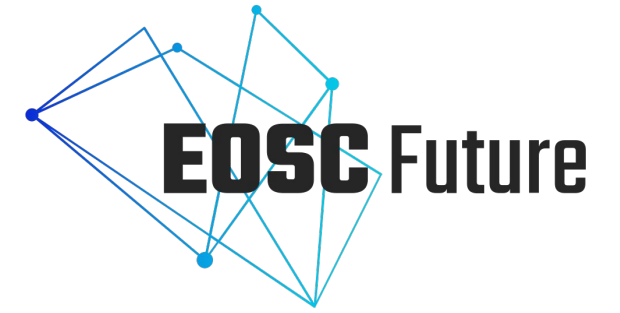


# Preprocessor Unit



# Preprocessor Unit

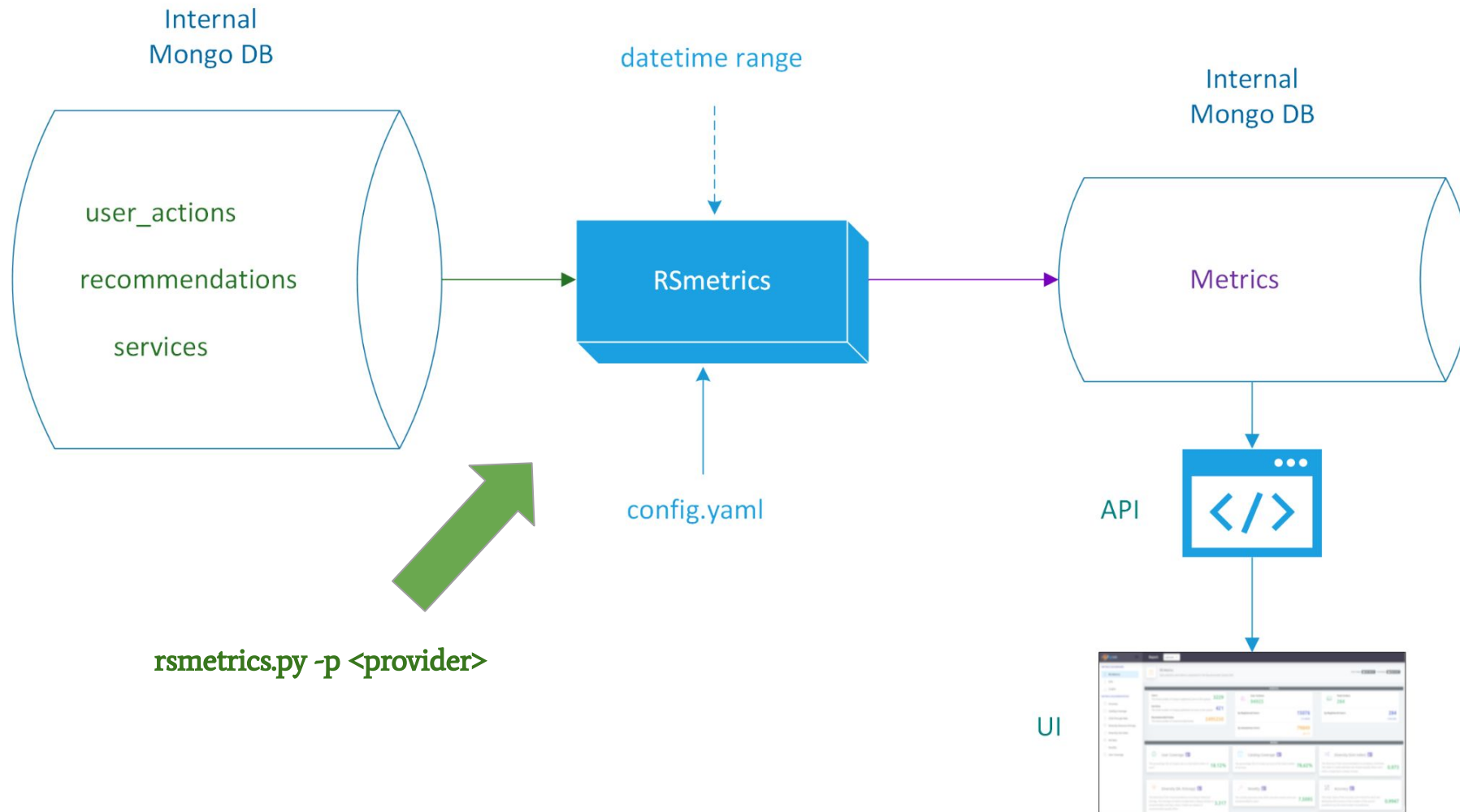




# RMF Components

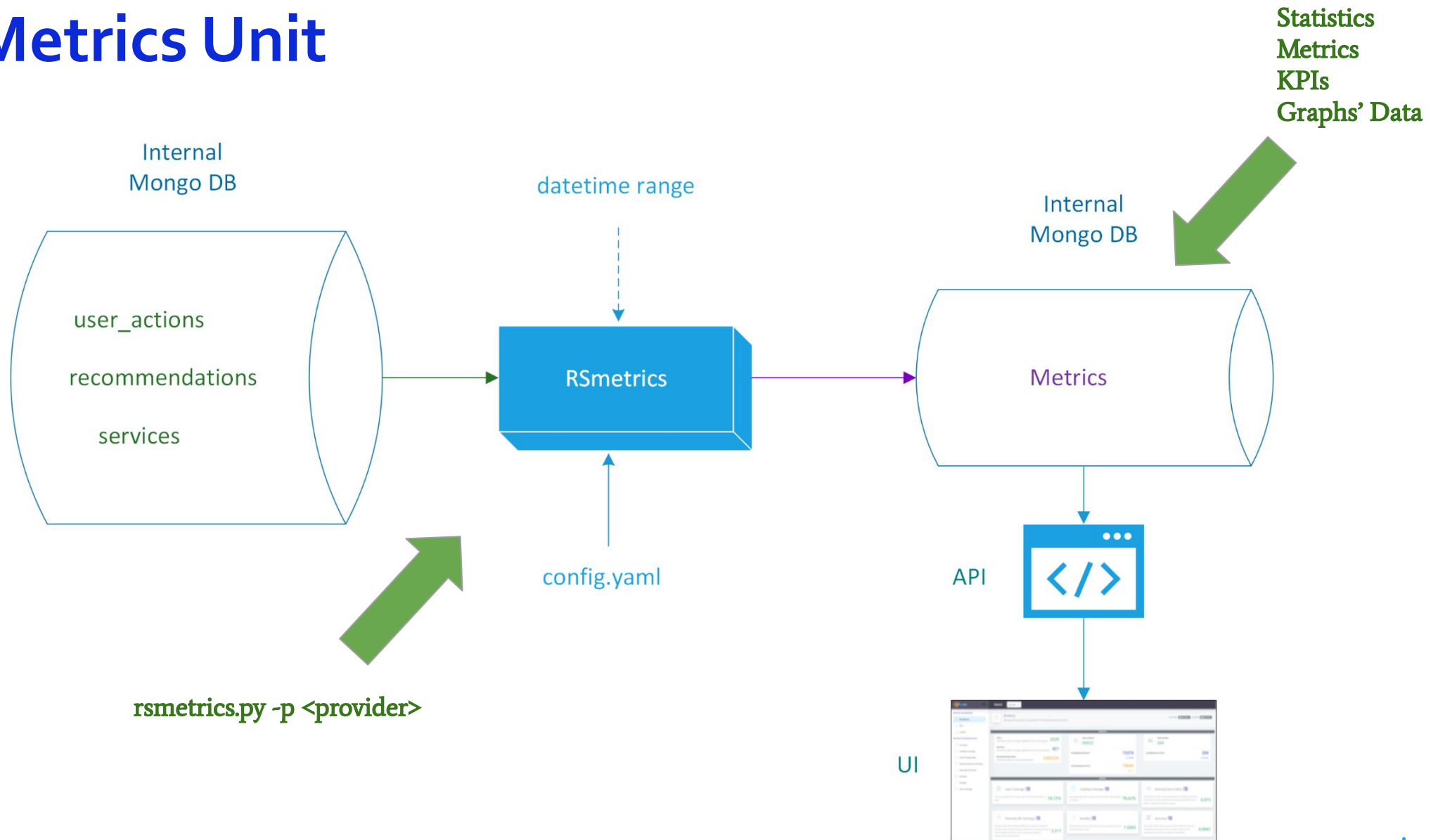
RS Metrics Unit

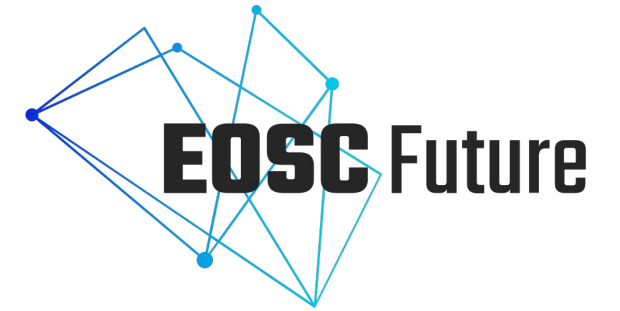
# RS Metrics Unit





# RS Metrics Unit

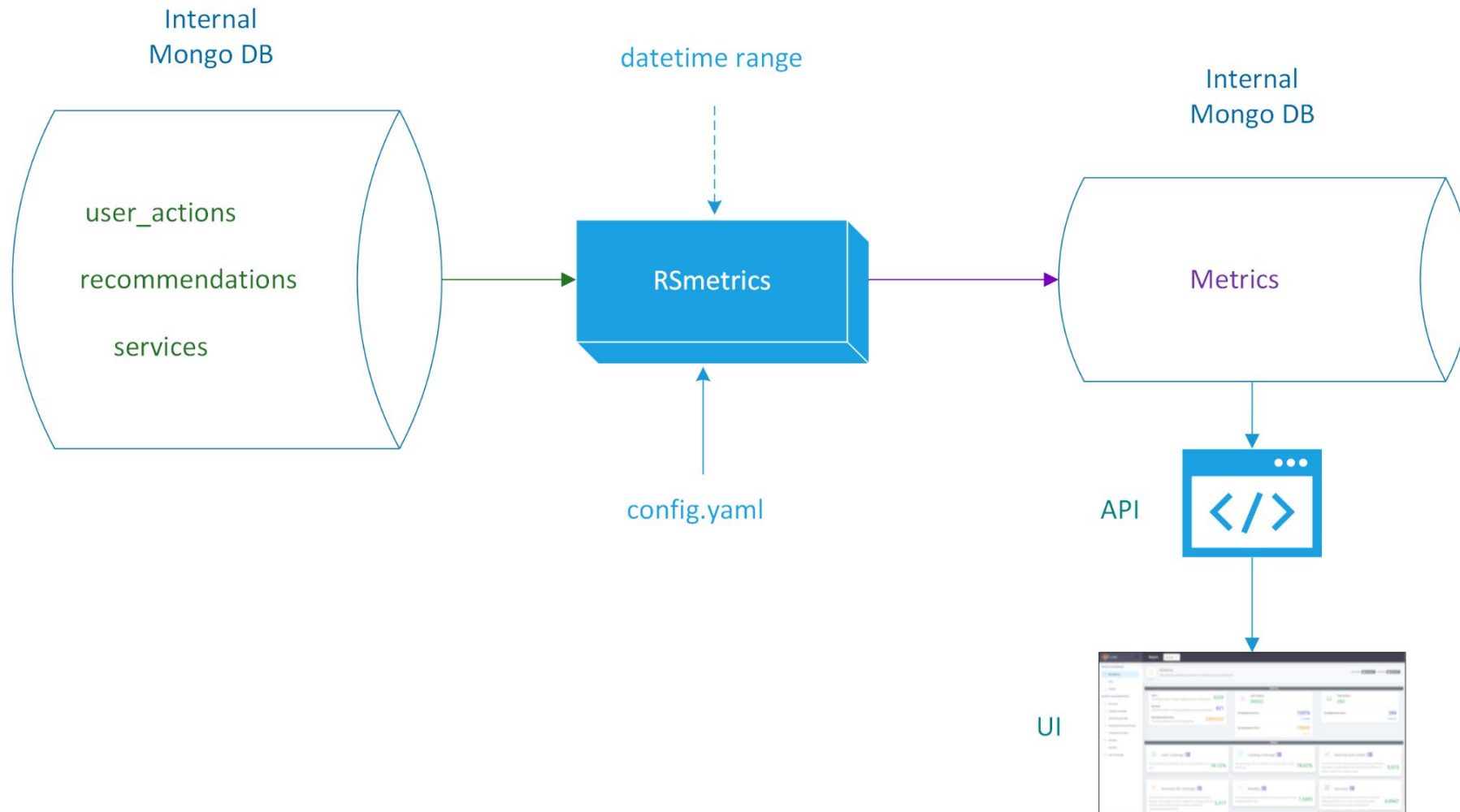




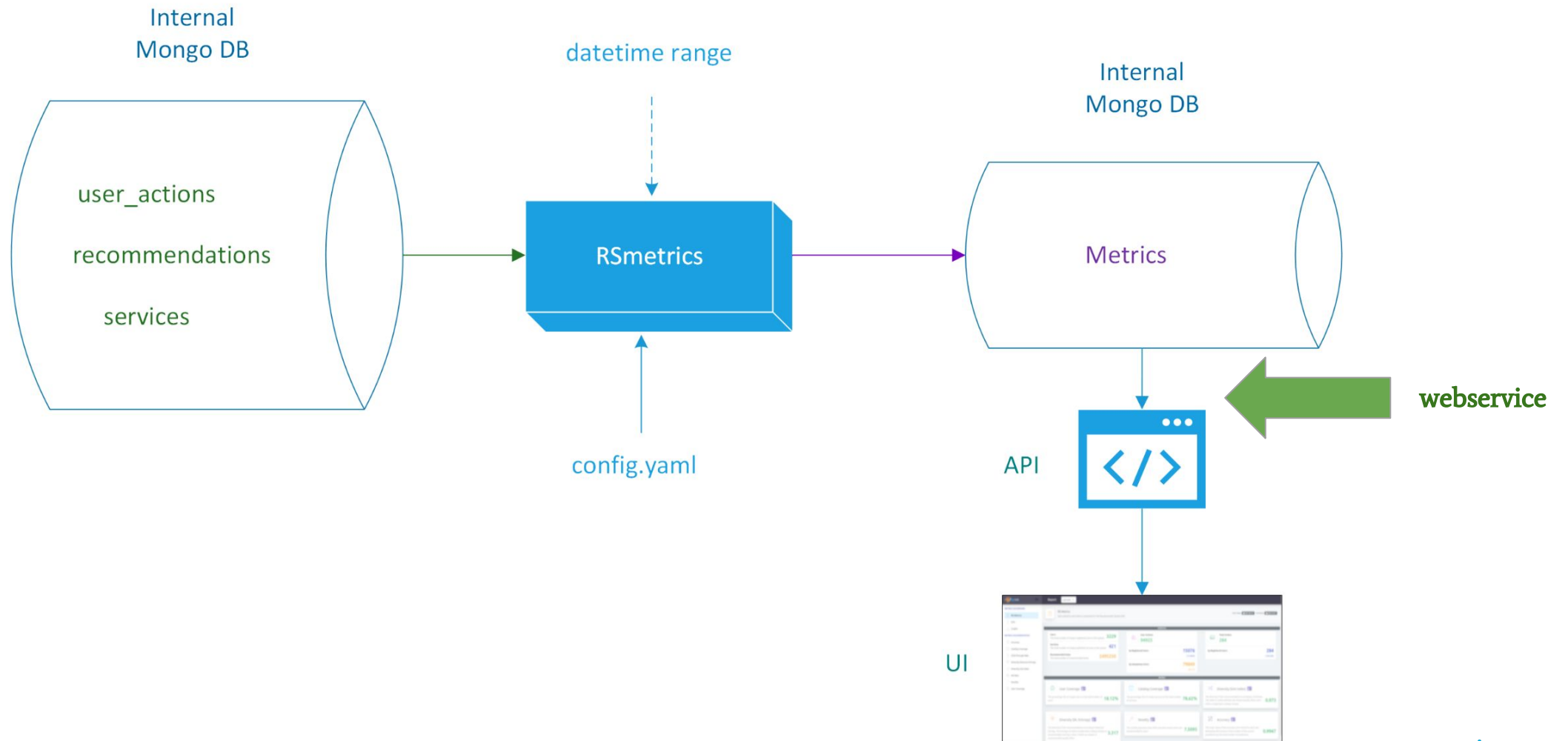
# RMF Components

Rest API/ UI Dashboard Unit

# REST API/UI Dashboard Unit



# REST API/UI Dashboard Unit





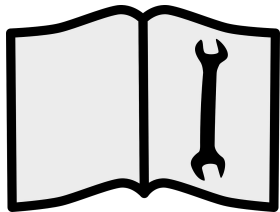
## More to see ...



[overview](#)



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

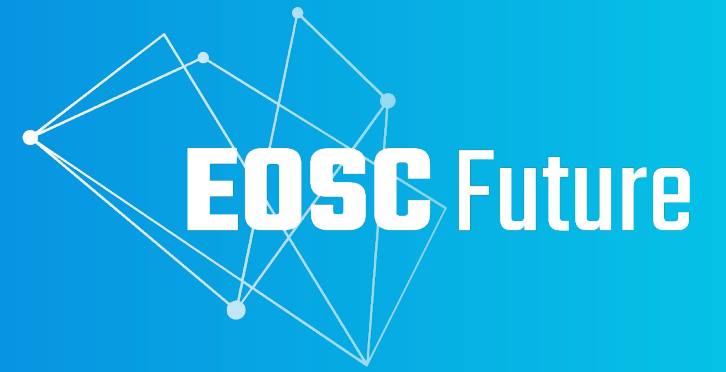


<https://argoeu.github.io/eosc-recommender-metrics/docs>



# Ready to answer your questions!





# Thank you for your attention

The EOSC Future project is co-funded by the  
European Union Horizon Programme call  
INFRAEOSC-03-2020, Grant Agreement 101017536

