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MSc Artificial Intelligence and Data Engineering

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Large Scale and Multi-Structured Databases

Weather Management Application

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Introduction

1.1 Application Goal

This application is designed to offer a comprehensive meteorological service that integrates both real-time **weather forecasting** and access to **historical weather** data for the most relevant and populous Italian cities.

The first functionality of the application is a standard, yet essential, weather forecasting service. This is crucial for all types of users the application may have, as it supports daily activity planning. For this component, we focused particularly on responsiveness and service availability. Alongside the forecasting service, we developed a module for the analysis of historical weather measurements, enabling accurate and detailed examinations of trends, specific events, and anomalies.

Building upon historical weather data, we included a dedicated section for particularly significant meteorological phenomena, referred to as **Extreme Weather Events** (sometimes abbreviated as EWE). These events can be crucial for identifying patterns and trends in climate behavior, providing a high-level overview without the need for in-depth analysis of individual measurements.

1.2 Big Data Analysis

Meteorological data is characterized by both high **volume**—due to the numerous variables involved—and high **velocity**, stemming from the frequency at which the data must be updated. These characteristics make meteorological data an ideal candidate for the application of various Big Data Analysis paradigms, which we employed in both the forecasting and historical measurement components of the system.

1.3 Code repository

This documentation describes the main ideas and strategies used to implement the application, along with an overview of the completed final product and some snapshot of the code. The entire code can be found in the repository:

https://github.com/AlessioSmn/WheaterAppServer/tree/vmCluster

Application Design Overview

2.1 Structure of the application

The application has been developed using **Java** programming language, following the **client-server** paradigm. The server side has been implemented with the **Spring** framework. It provides to the client a **set of APIs**, described in the following chapters, to access weather forecasts and historical data. The client side, even if not developed (as per specifications), is supposed to be composed of a graphical interface with a section dedicated to sign-up/login and another to display the data received from the server via the API.

2.2 Use-Case Diagram

In the use case diagram, three different actors are identified: **Guest**, **Registered User**, and **Admin**.

- Guest can exclusively view weather forecasts and historical data.
- Registered User can add any city to their list of *followed cities*, a feature particularly useful for enabling a personalized graphical interface and a client-side notification system.
- Admin has extended privileges and can trigger the replacement of outdated data with more recent information, as well as update thresholds, which are described in the next sub-section. Moreover, the admin can add cities in the database. This is a key feature, in fact, initially the application only provides the information of the Italian province capitals. This can be not enough to fully describe the weather situation in Italy, since some parts are really far from the province capitals. Some examples can be Elba Island or Abetone, which are both relatively far from their province capital (respectively, Livorno and Pistoia).

2.3 UML

The UML class diagram models the structure of a weather monitoring and forecasting system with three types of users: **Guest**, **Registered**, and **Admin**, all of which inherit from a common **User** class. **Guest** users have no credentials and are limited to viewing forecasts and historical data. **Registered** users, identified by a username, email, and password, can follow multiple cities, enabling features such as notifications and personalized interfaces. **Admins** have similar credentials but are responsible for managing the system, including updating data and threshold parameters. The core entity is **City**, characterized by attributes like name, region, geographic coordinates, elevation, and the number of followers. Each city is associated with:

- Multiple Predictions, which contain forecast data such as temperature, rainfall, snowfall, wind speed, and the number of hours in advance.
- Multiple HourlyMeasurement, which store historical weather observations with timestamps. Every observation consists in: temperature, rainfall, wind-speed and snowfall.

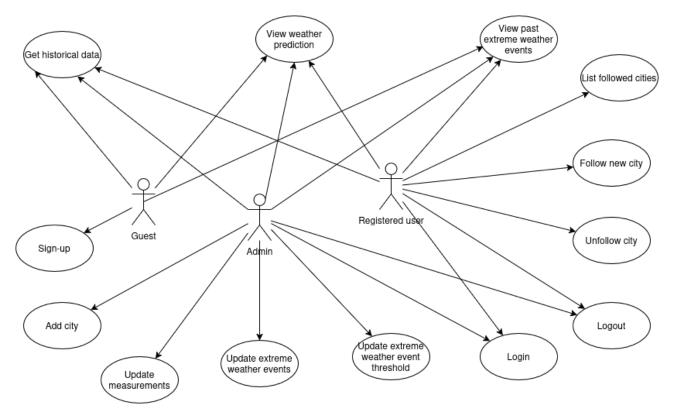


Figure 1: Main actors

- A single set of ExtremeWeatherThresholds, specifying maximum and minimum limits for temperature, rainfall, snowfall, and wind speed¹.
- Multiple ExtremeEvents, described by category (Rainfall, Snowstorm, ecc.), strength, and a date range.

2.4 Functional Requirements

A list of the main functional requirements of the application:

- Users can register, log-in and log-out.
- The application allows users to view forecasts weather conditions for the next few days for any city (both province capital and not).
- Authenticated users have access to advanced features, such as: saving favorite cities for quick access to related forecasts.
- The application stores historical measurements and records of extreme weather events: users can explore these historical data through built-in analytical tools.
- Both historical weather data and forecast weather data is obtained from external APIs and is updated regularly.
- Admin users can update parameters and add elements to the database in order to obtain a more precise service;

¹These thresholds are used to define whether or not a measurement is an extreme event. For example, a heavy snowfall in Aosta is not considered extreme as if the same amount of snow fell in Syracuse.

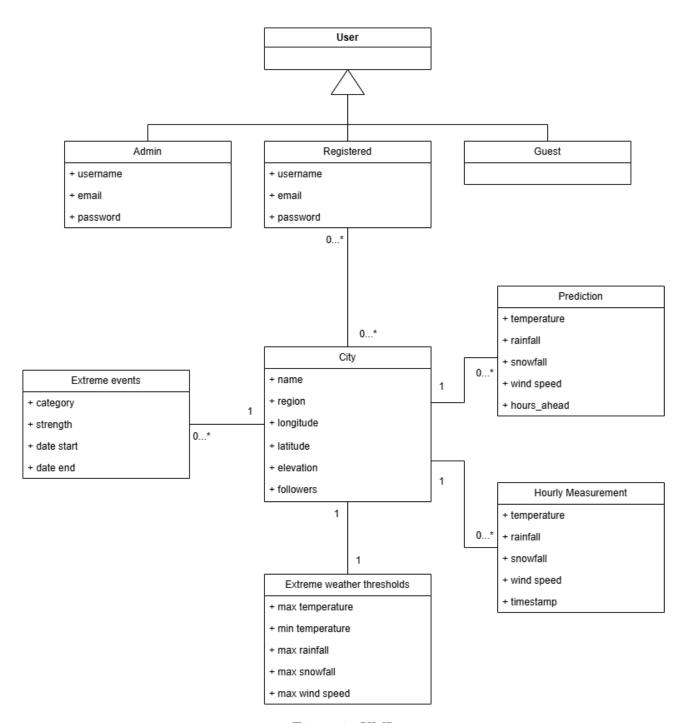


Figure 2: UML

2.5 Non-Functional Requirements

The primary non-functional requirement of the application is **performance**, with a specific focus on minimizing read time as much as possible. This requirement serves as the foundation for many of the implementation and data modeling decisions that will follow. In the NoSQL database context, performance translates not only into fast response times, but also into having a highly available service, capable of serving user requests with minimal latency even under heavy load. Ensuring fast access to data is critical, especially in scenarios where users expect instantaneous feedback or where large volumes of read operations are performed frequently. Additionally, an authentication mechanism is implemented to ensure secure access to the database. Upon login, the user receives a token, which must be included in the HTTP requests when accessing certain APIs. This token also serves to distinguish between Admin and Standard users.

2.6 API used to retrieving data

All the data stored in the database are retrieving from API-service called **Open Meteo**. The service provides both historical measurements from 1940 and forecast, filterable using data, timestamp, latitude and longitude. Moreover, data can be retrieving both using CSV and JSON. Our application uses JSON, since also MongoDB uses this format. Here's the APIs uri, respectively historical and forecast:

https://archive-api.open-meteo.com/v1/archive

https://api.open-meteo.com/v1/forecast

2.7 Databases architectures used

The NoSQL architecture used includes both a **DocumentDB** and a **Key-ValueDB**. The key-value store was chosen because the application requires fast reads and does not involve complex queries. Moreover, since forecast data changes frequently and must be updated or retrieved quickly, it is crucial to rely on a DBMS that supports high-performance access. On the other hand, the DocumentDB is well-suited for analytics purposes. For this reason, real-time forecast data is stored in the Key-Value DB, while historical records and extreme weather events are stored in the DocumentDB. Moreover, the DocumentDB stored the user information, since the this type of information is not too volatile. To identify a session is defined a token: this is stored within key-valueDB.

2.8 Some Mockups

These are very simple mockups that represent a possible GUI for our application. A registered user called *John Baker* is seeing Florence forecast, selecting a new city from the followed ones and searching for historical data.



Figure 3: Mockup - page 1

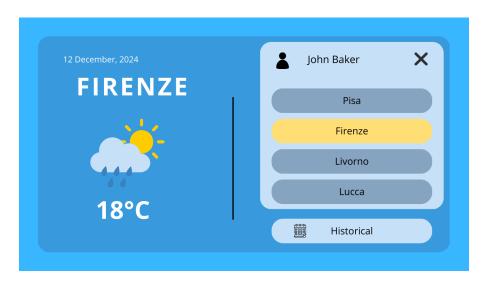


Figure 4: Mockup - page 2

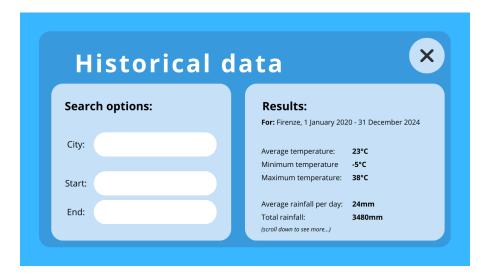


Figure 5: Mockup - page 3

Data Modeling and Indexes

3.1 MongoDB entities

The application uses MongoDB to store cities information, user information, historical weather data and extreme weather event entities. The data is organized across multiple collections, each with a specific purpose and structure.

3.1.1 Users Collection

User information is stored in a collection named users. Each document in this collection contains the following fields:

- username defined as unique
- password saved encrypted
- email defined as unique
- listCityId a list representing the cities followed by the user

The data modeling strategy adopted is *data linking*, as only the identifiers of the followed cities are needed—there is no requirement to include additional city information within the user documents.

3.1.2 Cities Collection

The cities collection holds information about each city. Every document includes:

- name
- region
- coordinates both latitude and longitude
- threshold a city-specific array weather threshold
- followers number of users that follow city: it can be useful from a client-side point of view
- city_id a unique identifier in the format: region-code_city-code_latitude_longitude
- eweList a list of the extreme weather event (ewe) that have affected the city. Every ewe has this structure
 - category: Indicates the type of event. The possible categories are defined in the ExtremeWeatherEventCategory enum.
 - strength: Represents the severity or intensity of the event.
 - dateStart: The timestamp indicating when the event was first detected.
 - dateEnd: The timestamp indicating when the event concluded.

In this case, data embedding was prefer in respect to than data linking. This is a (1:N) entity between city entity and ewe entity and access for the extreme weather event is performed mainly using the city_id. Since ewes are rare (of course, they are extreme), the size of the document would not be excessively large even if a dataset from the 80s is used.

3.1.3 Measurements Collection

Historical weather data is stored in a separate collection, where each document represents an hourly measurement. The fields include:

- temperature [C°]
- rainfall [mm/h]
- snowfall [mm/h]
- windspeed [km/h]
- timestamp in the form: YYYY-MM-DDTHH:MM:SS:sss.Z
- city_id linking the measurement to a specific city

The chosen data modeling strategy is *document linking*, rather than embedding. Embedding hourly measurements within each city document would have led to excessively large documents, approaching the MongoDB document size limit of 16MB. By separating measurements into their own collection and linking them via the city_id, the application achieves better scalability and maintainability.

3.2 Mongo Indexes

Within the DocumentDB, some indexes were defined in order to make reads faster. Since our application is more focused on reads, the side effects that the performance indicators of the writings have not been taken into account.

All the indexes has been defined in the Java class MongoInitiliazer.

3.2.1 User Indexes

Considering that the most frequent operation is certainly the login, an index has been defined on the **username** field.

3.2.2 City Indexes

Access to the city is largely done using the city_id, so the **default index** defined by MongoDB is sufficient.² This is fundamental since extreme weather events are performed using city-id as parameter.

3.2.3 Measurement Indexes

The analytics defined within the application analyze the measurements by city. Therefore, the index has been defined using cityId and time. This is certainly an expensive index (from a storage point of view), but it allows to have fast read of the hourly_measurements, a fundamental feature for the analytics.

²MongoDB provides a default index for the _id field.

3.3 Redis entities

Redis is used to store cities and their corresponding forecasts. City keys follow the pattern: city:{regioncode}-citycode-latitude-longitude,

e.g., $city: \{tus\} - pis - 43.7085 - 10.4036.$

These keys are essential for supporting the arbitrary city forecast feature (see next section), which requires retrieving cities by region and computing distances between them. The values retrievable using these keys is useful the information (region and elevation) used in the arbitrary city forecast functionality. Thus, the value-type used is **hash**.

Forecasts are stored using the format:

forecast:{regioncode}-citycode-latitude-longitude:date,

e.g., forecast: {tus}-pis-43.7085-10.4036:2025-06-08.

This key structure allows retrieving the entire daily forecast, consistent with the design of the external APIs, which operate on daily or weekly granularity. The value associated with each forecast key includes temperature, rainfall, snowfall, windspeed, and timestamp, matching the schema used for historical measurements. This is simply stored as a **string**.

3.4 Intra-DB consistency

The only common entity between the two databases is city. As a result, the city insertion operation is entirely **rolled back** if one or both write operations fail. Additional details are illustrated in the implementation.

Code Implementation

The main components of the server application, located in the src/main/java/it.unipi.lsmsd/folder, are the following:

- controller: handles HTTP requests and routes them to services
- service: contains business logic
- DTO: defines data transfer objects for external communication
- model: defines domain entities and data structures
- repository: manages data access and persistence
- utility: provides helper functions and common utilities

4.1 Controllers

The controller classes define the API endpoints of the application. They do not implement logic directly; they delegate all processing to the appropriate service classes. Controllers are responsible for handling HTTP requests, invoking the corresponding service methods, and managing errors by mapping them to appropriate HTTP status codes and response formats.

- UserController: Handles all user-related operations, such as registration, login, and logout.
- CityController: Handles all admin-reserved operations related to cities.
- FavoriteCitiesController: Handles all operations reserved for registered users related to their list of followed cities.
- EweController Extreme Weather Event Controller: Handles all admin-reserved operations related to extreme weather events.
- HourlyMeasurementController: Handles all operations related to the retrieval of forecasts.
- AnalyticsController: Handles all API calls related to analytics on both historical data and extreme weather events.
- DataManagerController: Handles all the admin-reserved API calls to update measurements and forecasts in the database. It is also to initialize the databases: this part is not intended for use during the normal operation of the application.
- GlobalExceptionHandler: Handles exceptions caused by invalid enum values passed as parameters. It is used by all APIs that take either an ExtremeWeatherEventCategory or a MeasurementField as input.

4.2 Service

The service package contains the classes in which the core logic of the application is implemented. Each controller delegates the execution of operations to a corresponding service class, which encapsulates the main logic and often interacts with the database through repository functions.

UserService Defines the functions login, logout, register, getUserFromToken, and check-Role. The last two functions are central to the implementation of authorization control for specific functionalities.

CityService Defines the functions saveCity and updateCityThresholds, which implement the logic for adding a new city to the database and updating the extreme weather event threshold values for a given city, respectively. Other functions include setLastEweUpdateById, get-LastEweUpdateById, and setLastMeasurementUpdateById, which serve as utility methods to support the refreshing of historical measurements and extreme weather events. All these functions interact with the database through repository methods. The function saveCity is fundamental in the application environment. It adds a city in Mongo and Redis (if only one write fails, everything is rolled back), in order to obtain a better prediction and historical service.

FavoriteService Defines the functions getFavorites, addToFavorites and removeFrom-Favorites, which manage the list of followed cities for each user. Both functions rely on getUserFromToken from UserService to identify the requesting user based on the authorization token. The last two functions are annotated with @Transactional to ensure atomicity of two operations: updating listCityId and followers.

ExtremeWeatherEventService Defines the main functions updateExtremeWeatherEvent and cleanExtremeWeatherEventDuplicates. Additional functions include updateExtremeWeatherEventAll, updateExtremeWeatherEventAutomatic, cleanExtremeWeatherEventDuplicates-Range, and cleanExtremeWeatherEventDuplicatesAll, which internally invoke the main methods with different time range scopes and are the ones directly called by the controllers.

- updateExtremeWeatherEvent (Figures 6 and 7) This function takes as input a target city and a time range. It retrieves all measurements within the specified range and the corresponding extreme weather event thresholds for the given city. It then compares the measurements against the thresholds: when an out-of-threshold value is detected, a new extreme weather event entity is created. The entity may be further updated in terms of strength and duration if subsequent out-of-threshold measurements are found.
- cleanExtremeWeatherEventDuplicates Since the updateExtremeWeatherEvent function does not check for pre-existing overlapping extreme weather events, this method is designed to clean up potential duplicates by removing entities that overlap in time.

```
public List-ExtremeMeatherEvent> updateExtremeMeatherEvent( 2 usages 4 Simported 1
    String cityId
    LocalDateTime startImeInterval,
    LocalDateTime endTimeInterval
) throws CityNotFoundException, ThresholdsNotPresentException {

// Gets all measurements for a given city
    Date endTime = Date.from(startTimeInterval.toInstant(ZoneOffset.UTC));
    Date endTime = Date.from(cendTimeInterval.toInstant(ZoneOffset.UTC));
    List-HourlyMeasurement> hourlyMeasurements = hourlyMeasurementRepository.findByCityIdAndTimeBetweenOrderByTimeTimeAsc(cityId, startTime, endTime);

// Gets the target city, in order to get the thresholds
    Optional-Cityy city = cityRepository.findById(cityId);
    if (city.isEpty(O)
        throw new CityNotFoundException("Specified city + cityId + " was not found");

// Check that the city has all the necessary thresholds
    if*(!hosCityLtIThresholdsFields(city.get(O))
        throw new ThresholdsNotPresentException("City * cityId + " doesn't have all threshold fields correctly specified");

// Retrieve the city's extreme weather event thresholds
    EWEIhneshold exclmresholds = city.get(O.getEmeThresholds();

// Array of ENE found and engoing ENES
    List-QuadrupleEWEInformationHolder> ongoingExtremeWeatherEvents = new ArrayList<-(getEmptyListOffENES());

// Morking array of found and engoing ENES
    List-QuadrupleEWEInformationHolder> currentLocalENEs = getListOfCurrentLocalENEs(city.get(O), startTimeInterval);

// Putting the currentLocalENEs into the ongoingExtremeWeatherEvents array
    for (QuadrupLeEWEInformationHolder dbENE : currentLocalENEs) {
    int index = dbENE.getCategory().ordinal();
    ongoingExtremeWeatherEvents.set(index, dbENE);
}
```

Figure 6: updateExtremeWeatherEvent function - part 1

```
// List of completed EWEs, to be returned
List-ExtremeWeatherEvents completedEWEs = new ArrayList<>();

// Checks all measurements against all thresholds
for (HourlyMeasurement measurement : hourlyMeasurements) {

// Gets the list of current EWE
foundEWEs = getCurrentEWEs(measurement, eweThresholds);

// Loop over all EWE categories using index,
// thanks to the fact that the two arrays are always ordinated by category in the same way
for (int i = 0; i < ExtremeWeatherEventCategory.values().length; i++) {

// Find the ongoing ewe strength and found ewe strength
int foundStrength = foundEWEs.get(1).getSecond();
int ongoingStrength = ongoingExtremeWeatherEvents.get(1).getStrength();

// Note: strength == 0 means no ewe is found, the measurement value is under the threshold limit

// Set new startDate if a new EWE is found
if (foundStrength > 0 && ongoingStrength == 0)[...]

// Update ongoing event strength if a greater strength is detected
if (foundStrength > ongoingStrength && ongoingStrength > 0) [...]

// If the found strength is 0 but the event was ongoing, it means the event has ended
if (foundStrength == 0 && ongoingStrength > 0) [...]

}

// Check for still ongoing EWE, save them to DB in the case
for (QuadrupLeEWEInformationHolder ongoingGwe : ongoingExtremeWeatherEvents)[...]

// Update the city, with all new EWEs inserted in the list
cityRepository.save(city.get());

// Returns a list of found and inserted EWEs
return compltedEWEs;
}
```

Figure 7: updateExtremeWeatherEvent function - part 2

HourlyMeasurementService Defines the functions saveHourlyMeasurements, used during initialization to insert all historical measurements from a JSON file into MongoDB, and refreshHourlyMeasurementsAutomaticFromOpenMeteo (Figure 8), a function designed to be called periodically to retrieve the latest measurements from OpenMeteo and store them in the database.

```
public void refreshHourlyMeasurementsAutomaticFromOpenMeteo(String cityId) throws JsonProcessingException {
   Optional<City> optionalCity = cityRepository.findById(cityId);
   if(optionalCity.isEmpty()){
   City city = optionalCity.get();
   LocalDateTime lastMeasurementUpdate = city.getLastMeasurementUpdate();
   LocalDate lastMeasurementUpdate_date = lastMeasurementUpdate.toLocalDate().plusDays( daysToAdd: 1);
   LocalDate yesterday = LocalDate.now().minusDays( daysToSubtract: 1);
   APIResponseDTO responseDTO = dataHarvestService.qetCityHistoricalMeasurement(
           city.getLatitude(),
           city.getLongitude(),
           lastMeasurementUpdate_date.toString(),
           yesterday.toString()
   HourlyMeasurementDTO hourlyMeasurementDTO = responseDTO.getHourly();
   hourlyMeasurementDTO.setCityId(cityId);
   saveHourlyMeasurements(hourlyMeasurementDTO);
   cityService.setLastMeasurementUpdateById(cityId, yesterday.atTime( hour: 23, minute: 0));
```

Figure 8: refreshMeasurementsFromOpenMeteo function

RedisForecastService This class defines all functions related to the management of weather forecasts. The function refreshForecastAutomaticFromOpenMeteo (Figure 9) is responsible for retrieving updated forecasts from OpenMeteo and saving them into Redis via the saveForecast function. The method deleteAllForecast is used to remove existing forecasts before inserting new data into Redis.

Figure 9: refreshForecastFromOpenMeteo function

The functions getForecastTargetDay (Figure 10), get7DayForecast, and getForecastAr-bitraryCityTargetDay implements the logic required to handle requests from controllers and to retrieve or estimate weather forecasts for an arbitrary location specified by the client. The method takes the following input parameters: region, latitude, longitude, and elevation. To estimate the forecast at the target location, the system first retrieves the city_ids of all cities within the specified region, as stored in the database. Then, it computes a weighted average of the weather values (e.g., temperature, rainfall) retrieved from the Forecast entity. The weights used in the average are based on the inverse of a parameter called effective-distance which is calculate with the difference of the distance between the target location and each city and the elevation difference³. The distance is calculated using the Haversine algorithm, which relies on latitude and longitude values. Elevation data for the cities is retrieved from Redis by accessing the City entity using the corresponding city_ids.

³Both the distance and the elevation difference are scaled by two empirical factors, called *alpha* and *beta*, which are determined by comparing the output of the method with actual forecasts obtained from some websites.

Figure 10: getForecastTargetDay function

AnalyticsService Defines all the functions related to analytics on historical measurements. A large number of methods are implemented — one for each corresponding controller endpoint— but they all share a common structure: they receive query parameters as input and perform the computation using a MongoDB aggregation pipeline.

We take this function (Figure 11) as example: it retrieves the lowest measurement values for a specified city and measurement type within a given time range. It returns the numMeasurementsToFind lowest entries by applying a MongoDB aggregation pipeline that filters by city and date, sorts by the selected field, and adds to the result the formatted date and time fields.

```
public List<Document> lowestMeasurementsInCityDuringPeriod( 1usage ᆂ Simocnini
       String cityId,
       MeasurementField measurementField,
       LocalDateTime startDate.
       LocalDateTime endDate,
       int numMeasurementsToFind
   String fieldName = getFieldName(measurementField);
   return StreamSupport.stream(measurementCollection.aggregate(
           Arrays.asList(
                    match(and(
                            eq( fieldName: "cityId", cityId),
                            gte( fieldName: "time", startDate),
                            lte( fieldName: "time", endDate)
                    sort(orderBy(ascending(fieldName))),
                    limit(numMeasurementsToFind),
                    addFields(
                            new Field<>( name: "day", new Document("$dateToString", new Document()
                                     .append("format", "%Y-%m-%d")
                                     .append("date", "$time"))),
                            new Field<>( name: "timeOfDay", new Document("$dateToString", new Document()
                                    .append("format", "%H:%M:%S")
                                     .append("date", "$time")))
                    project(fields(
                            include( ...fieldNames: "cityId", "day", "timeOfDay", fieldName)
            )).spliterator(), parallel: false)
            .collect(Collectors.toList());
```

Figure 11: lowestMeasurementsInCityDuringPeriod function

AutomatingService Defines the functions updateMeasurements, updateExtremeWeatherEvents, and updateForecasts. The purpose of this service is to automate the daily tasks that would otherwise need to be performed manually for each city by an administrator. The main update routine is designed to be executed daily when the system load is estimated to be at its lowest. These functions are invoked by the /update endpoints of the DataManagerController.

- updateMeasurements Retrieves from OpenMeteo the measurements of the previous day for all cities stored in the database and inserts them into the measurements collection. It calls the function shown in Figure 8 for all the cities in the database.
- updateExtremeWeatherEvents Calculates any new extreme weather events or updates existing ones based on the newly acquired measurements.
- updateForecasts Retrieves from OpenMeteo the forecasts for the upcoming week for all cities stored in the database and inserts them into Redis. It calls the function shown in Figure 9 over all the cities in the database.

4.3 DTO

The DTO (Data Transfer Object) folder includes classes used to encapsulate and transfer data between the different layers of the application. These classes are particularly useful as they allow the inclusion of additional fields or transformations without modifying the structure of the model classes. For this reason, they serve as flexible data carriers tailored to specific use cases such as API responses or service outputs. A detailed discussion of each DTO is omitted, as they concern purely implementation-level aspects.

4.4 Model

In the model folder are defined all the classes representing the core entities of the application.

- User
- City
- ExtremeWeatherEvent
- HourlyMeasurement

This folder also contains enum types, used to standardize and constrain values across the application:

- Role Defines the possible roles of a user: ADMIN or REGISTERED_USER.
- ExtremeWeatherEventCategory Defines the possible categories of extreme weather events: RAINSTORM, SNOWSTORM, HURRICANE, HEATWAVE, COLDWAVE.
- MeasurementField Used in analytics APIs to specify a target measurement; possible values are: RAINFALL, SNOWFALL, WINDSPEED, TEMPERATURE.

4.5 Repository

The repository folder contains all the interfaces responsible for interacting with the MongoDB database. Each repository extends the MongoRepository interface and provides access to the corresponding collection. These interfaces define the methods used throughout the various service classes to access data, mainly defining standard operations such as findById. The defined interfaces and relative methods are: The following interfaces are defined to handle data persistence:

- UserRepository manages user data access.
 - Optional < User > find By Username (String username);
 Retrieves a user by their username.
- CityRepository manages access to city-related data.
 - List<City> findAllByName(String name);
 Returns all cities with the given name.
 - List<CityBasicProjection> findAllBy();
 Returns all cities using the CityBasicProjection interface.
 - List<CityBasicProjection> findAllByNameOrderByFollowers(String name); Returns cities with the specified name, ordered by follower count.
 - Optional<City> findCityWithOngoingEvents(String cityId);
 Returns a city only if it has at least one ongoing event (i.e., dateEnd = null).
 - Optional<City> findCityWithEventsOfCategory(String cityId, String category);
 - Returns a city only if it contains at least one event of the specified category.

- Optional<City> findCityWithEventsStartedBefore(String cityId, LocalDate-Time startTime);
 - Returns a city only if it has events that started before the specified time.
- Optional<City> findCityWithEventsInRangeAndCategory(String cityId, String category, LocalDateTime start, LocalDateTime end);
 Returns a city only if it has events in the given time range and category.
- Hourly Measurement Repository handles hourly measurement data.
 - List<HourlyMeasurement> findByCityIdAndTimeBetweenOrderByTimeTimeAsc(String cityId, Date startTime, Date endTime);
 Returns all measurements for a given city within a time range, sorted by time.
 - Optional<HourlyMeasurement> findFirstByCityIdOrderByTimeAsc(String cityId);
 Returns the earliest recorded measurement for a given city.
 - void deleteByCityIdAndTimeBetween(String cityId, Date startTime, Date end-Time);
 - Deletes all measurements for a city within the specified time range.

Clustering and Sharding

To deploy our database into a distributed environment, we have access to 3 UniPi virtual machines: 10.1.1.9, 10.1.1.84 and 10.1.1.87.

Since the application is heavily **read-based** the **master-slave** architecture is by far the best option. The "standard" master-slave architecture would not provide **sharding**. However, completely abandoning sharding would be problematic, since the scalability relative to the amount of data would be compromised. To address this, the adopted solution combines both paradigms by assigning a dedicated master and replica to each shard — a pattern that corresponds to the default architecture of a Redis cluster.

5.1 MongoDB clustering

The mongoDB replicas have been configured using the following parameters:

```
rs.initiate({
    _id: "rswa",
    members: [
        { _id: 0, host: "10.1.1.87:27017", priority: 5 },
        { _id: 1, host: "10.1.1.84:27017", priority: 3 },
        { _id: 2, host: "10.1.1.9:27017", priority: 1 }
    ]
});
```

The connection was configured with w = 1 to ensure write acknowledgment from at least one node, thus ensuring **availability**, and readPreference = local to allow fast reads from the **nearest** node.

5.2 MongoDB sharding

The parameters shown in the previous subsection do **not** actually involve sharding. A possible design for sharding could be the following:

```
user: shard by _id;city: shard by _id;
```

hourly_measurement: shard by cityId;

This sharding structure allows you to perform analytics without pulling data from multiple nodes, since hourly measurement from the same city are stored in the **same physical machine**.

5.3 Redis Replica

Each node runs a Redis master and a Redis slave. The configuration is really simple:

```
port 6379 # for the slave, the port used is 6380
cluster-enabled yes
cluster-config-file nodes.conf
```

cluster-node-timeout 5000 appendonly yes bind 0.0.0.0 maxmemory 512mb maxmemory-policy allkeys-lfu

5.4 Redis Sharding

By default, Redis uses the **CRC-16** algorithm to calculate which hashslot a given key belongs to. Unless otherwise specified, this algorithm is performed on whole keys. Using {}, you can specify a substring of the key on which to perform CRC-16. Choosing the region-code as a substring results in data sharding that allows for fast operations, since you only need to access a single node to obtain forecast information for the same city/region.

Here some example regarding the city Pisa:

region: {tus}

city:{tus}-pis-43.7085-10.4036

forecast:{tus}-pis-43.7085-10.4036:2025-06-08

5.5 CAP theorem

The application is designed to be **AP**. While partition tolerance is essential, it is preferable to provide users with an always-available weather service rather than a strictly consistent one. Therefore, MongoDB parameters are configured to favor fast reads and high availability, as described in the previous section. Redis, by design, also operates as an AP system by default.

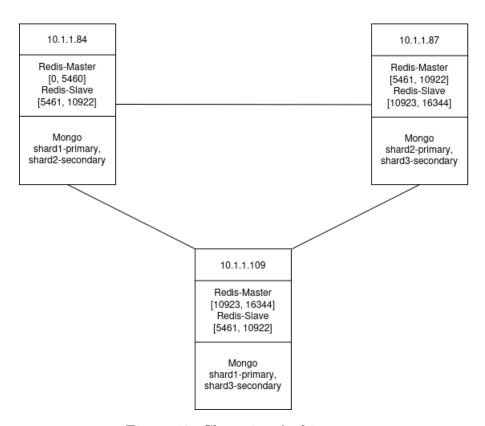


Figure 12: Clustering Architecture

API List

This is the list of APIs in the application. Some APIs are reserved for Admin users, others for Registered users.

6.1 Extreme Weather Event

DELETE /ewe/duplicates/range — Deletes duplicate Extreme Weather Events in the given time range, for the given city.

```
• cityId: string
```

• startTime: string

• endTime: string

DELETE /ewe/duplicates/all — Deletes duplicate Extreme Weather Events for the given city.

• cityId: string

6.2 User

POST /user/register — Lets a new user register to the application

• UserDTO: UserDTO

POST /user/login — Lets a registered user or an admin to login into the application

• UserDTO: UserDTO

POST /user/logout — Lets a registered user or an admin to logout into the application The UserDTO is structured as:

```
"username": "string",
   "password": "string",
   "email": "string",
   "role": "ADMIN or REGISTERED_USER",
   "adminCode": "string"
}
```

6.3 Forecasts

GET /forecast/today — Gets the forecast for the target city for the current day

• cityId: string

GET /forecast/today/arbitrary-city — Gets the forecast for any arbitrary city, not saved in the database, for the current day.

```
• region: string
   • latitude: number
   • longitude: number
   • elevation: number
GET /forecast/day — Gets the forecast for the target city for the target day
   • cityId: string
   • targetDate: string
GET /forecast/day/arbitrary-city — Gets the forecast for any arbitrary city, not saved
in the database, for the target day
   • region: string
   • latitude: number
   • longitude: number
   • elevation: number
   • targetDate: string
GET /forecast/week — Gets the forecast for the target city for the next seven days
   • cityId: string
6.4 City
POST /city/add — Inserts a new city into the application.
   • CityDTO: CityDTO
Where the cityDTO is structured as:
    "name": string,
    "latitude": number,
    "longitude": number,
    "elevation": number,
    "eweThresholds": {
        "rainfall": number,
```

{

}

"snowfall": number,

"windSpeed": number

"region": string

"maxTemperature": number, "minTemperature": number, **PUT /city/update-thresholds** — Updates the Extreme Weather Event threshold values for the specified city.

• CityDTO: CityDTO

GET /city/by-name — Retrieves information on all cities with the specified name.

• cityName: string

GET /city/all — Retrieves information on all cities stored in the database.

GET /favorites — Retrieves a list of the user favorite cities in the form: (name, region).

PUT /favorites — Allows a user to add a city to their list of followed cities.

• targetCityId: string

DELETE /favorites — Allows a user to remove a city from their list of followed cities.

• targetCityId: string

6.5 Admin periodic update

PUT /data-manager/update/forecasts — Updates the forecasts for the next seven days for all cities.

PUT /data-manager/update/measurements — Updates the measurements of all cities with the data from the previous day.

PUT /data-manager/update/ewes — Updates the extreme weather events for all cities based on the latest measurements.

6.6 Analytics on measurements per city

All these APIs take as arguments the target city, a measurement type, and a time range within which to search for the measurements.

GET /analytics/measurement — Gets a list of measurements of the given measurement type in the time range.

• cityId: string

• measurementField: string

• startDate: string

• endDate: string

 $\begin{tabular}{ll} {\tt GET /analytics/measurement/city/lowest} & $-$ Gets the lowest measurements in the time range. \\ \end{tabular}$

• cityId: string

• measurementField: string

• startDate: string

• endDate: string

GET /analytics/measurement/city/highest — Gets the highest measurements in the time range.

• cityId: string

• measurementField: string

• startDate: string

• endDate: string

GET /analytics/measurement/city/average — Gets the average measurement value in the time range.

• cityId: string

• measurementField: string

• startDate: string

• endDate: string

GET /analytics/measurement/city/average-per-month — Gets the average measurement value per month in the time range.

• cityId: string

• measurementField: string

• startDate: string

• endDate: string

6.7 Analytics on measurements per region

All these APIs take as arguments the target region, a measurement type, and a time range within which to search for the measurements.

GET /analytics/measurement/region/average — Gets the average values of the given measurement type for each city in the given region.

• region: string

• measurementField: string

• startDate: string

• endDate: string

GET /analytics/measurement/region/lowest — Gets the lowest value of the given measurement type recorded for each city in the given region.

• region: string

• measurementField: string

• startDate: string

• endDate: string

GET /analytics/measurement/region/highest — Gets the highest value of the given measurement type recorded for each city in the given region.

• region: string

• measurementField: string

• startDate: string

• endDate: string

6.8 Analytics on recent measurements

All these APIs take as arguments the number of days prior to the current day and a measurement type.

GET /analytics/measurement/recent/city/total-per-day — Gets the total measurement, calculated as the sum of the hourly measurements, of the target city for each day. ⁴

• cityId: string

• measurementField: string

• pastDays: number

GET /analytics/measurement/recent/city/average-per-day — Gets the average measurement of the target city for each day.

• cityId: string

• measurementField: string

• pastDays: number

GET /analytics/measurement/recent/region/total — Gets the total measurement over the past days of each city of the target region.

• region: string

• measurementField: string

• pastDays: number

⁴This API make sense only with a cumulative measurement, such as rainfall and snowfall.

GET /analytics/measurement/recent/region/average — Gets the average measurement over the past days of each city of the target region.

• region: string

• measurementField: string

• pastDays: number

6.9 Analytics on Extreme Weather Events

All these APIs take as argument an extreme weather event category.

GET /analytics/ewe/strength/maximum — Gets the maximum recorded strength of extreme weather events in the given time range for each city in the target city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/strength/average — Gets the average strength of extreme weather events in the given time range for each city in the target city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/duration/longest — Gets the maximum recorded duration, expressed in hours, of extreme weather events in the given time range for each city in the target city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/duration/average — Gets the average duration, expressed in hours, of extreme weather events in the given time range for each city in the target city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/count — Gets the number of extreme weather events in the time range for all cities within the specified city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/count-per-month — Gets the number of extreme weather events per month of the target city.

• extremeWeatherEventCategory: string

• cityId: string

• startDate: string

• endDate: string

GET /analytics/ewe/count-of-at-least-strength — Gets the number of extreme weather events, for all cities within the specified city, whose strength is greater than or equal to the given value.

• extremeWeatherEventCategory: string

• cityId: string

• minimumStrength: number

• startDate: string

• endDate: string