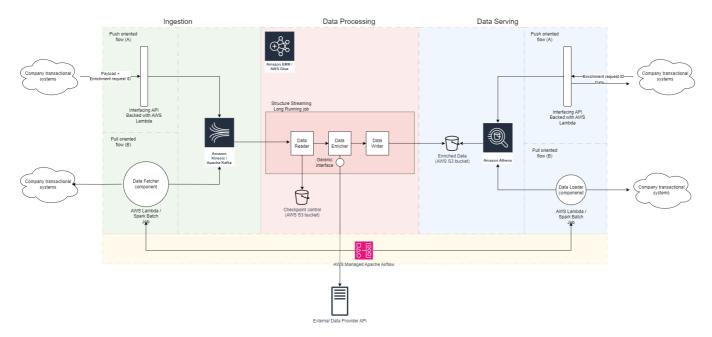
Data Enrichment Service Achitecture

This document aims to demonstrate the proposed design for the Data Enrichment Service and its components. In order to facilitate the reading, let us name the service as DES.

Solution design

The proposed long-term/ideal architecture diagram for the service described in the Take Home Assesment requirements is defined below:



Now let's dive in more details in each component of the proposed solution, in order to understand what role each one plays on it and how they interact with each other.

Ingestion layer

In order to start the DES ingestion layer, let us break it in two ways for inputting data into the service: (1) Push-oriented and (2) Pull-oriented methods. This definition takes into account that each company transaction service might have it's own level of maturity/complexity, therefore we need to provide a flexible integration as possible for others take advantage of our services provided.

Push-oriented

This approach is triggered by the transactional service that wants to enrich its data. For that, it was defined a interfacing API component on which this service can talk with the DES starting the enrichment request; it needs to send a JSON containing both the payload (contact list and a list of fields to be enriched), a **enrichment resquest ID (ERID)** and the total **number_of_records**. Here is an example:

```
POST https://des.internal.sayprimer.com/v2/enriched_data {
    "erid": "3c3f4b80-f728-4d7c-8567-1bc01df5f0a1",
    "contact_list": [
```

```
"first_name": "John",
            "last_name": "Doe",
            "company_name": "sayprimer.com"
        },
        {
            "first_name": "Foo",
            "last name": "Bar",
            "company_name": "sayprimer.com"
        },
        . . .
    ],
    "fields_to_be_enriched": [
        "professional_email",
        "phone number"
    ],
    "number_of_records": 100000
}
```

The erid is an important part of the flow because it defines a single enrichment transaction and as long as the requests made to the DES interfacing API have the same erid, they will be grouped at the output, in a single data object that can be referred to later.

The number_of_recrds should be used in the subsequent layers in order to control whether a particular Data Enrichment Transaction has finished or not.

It's important to highlight that it's an asynchronous request, therefore no enriched response will be returned at this point. Each request containing thousands of data objects will be forwarded to a Streaming Broker/Messaging System, so they can be processed and written in a near real-time fashion. Please check for more details in Processing Layer and Processing Layer.

P.S: Even though the diagram doesn't show the authentication on the API side, we can leverage some functionalities on AWS API Gateway for that purpose keeping our API secure and compliance.

Pull-oriented

In this method, the service can't, for any reason, integrate with our Asynchronous Ingest API but still, it needs to enrich data. The proposed architecture contemplates this situation as well, where instead of receiving the data enrichment requests, the DES service will start the process.

To accomplish this requirement, there are two key components of the architecture:

- A workflow orchestrator, AWS MWAA (Managed Workflow with Apache Airflow), accountable for managing the flows, job schedules, and dependencies for the data-pulling process
- An Data Fetcher component in a form of an ETL (read more on Config-based ETL framework), built on top of AWS Lambda or Spark batch jobs, that can handle multiple data sources using generic interfaces and specialized mechanism source-oriented implemented using object inheritance (circles on the architecture).

Either the service has to query the transactional data directly in their database, or APIs or has to read from an intermediate storage layer (s3 buckets, FTP serves, and others), by using the pull-oriented method, the service will also be able to bring the data to the service and process it in a scalable fashion. After querying the data on the source, it will forward it to the Broker/Messaging system that keeps the boundaries between the ingestion and processing layer.

One important aspect to highlight is that, even though there is no external service starting the data enrichment transaction, the Data Fetcher component must also generate a erid field so we can keep consistent on the data serving layer. In order to mediate the conversation between the Data Fetcher and the Data Loader components (see Serving data in a Pull-oriented flow), we can leverage AWS Managed Airflow in order to hold the erid for a particular ETL execution.

Broker/Messaging system

It will be one of the key components of the DES architecture that will provide us high scalability, as well as a fault-tolerant service which linked to the processing layer will form the high performatic core of the solution.

The broker will act as a data storage engine for real-time data reading, allowing streaming processes to read from it in a performant way, at the same time ensuring the data consistency with an *end-to-end exactly-once* semantics in case of failure.

Processing Layer

This layer is the DES functionality core, accountable for performing the data enrichment in fact. The approach chosen was streaming given the requirements for high performance and fault-tolerance; the suggested streaming job can be built on top of AWS EMR or AWS Glue framework. The streaming job logic will consist of three main internal "controllers":

- Data reader: process in charge of reading the data from the Broker as well as managing the checkpointing of the solution, so in case of failures, the solution won't either generate data duplication or data holes. It will start exactly from where it was stopped (using the offset concept)
- Data Enricher: it's the process that will perform the enrichment of the data, using external APIs for this purpose. Therefore it needs to control the connection to the N Data Source APIs implemented, the data rate that will be sent in each request, the number of requests, and other configurations (see more details below)
- Data Writer: it's a process in charge of writing the data in micro-batches to the output storage (s3 bucket), allowing the enriched data to be served.

Data Enricher

Interactions with APIs such as authentication, pagination, and endpoint convention might employ different techniques/frameworks, but the overall workflow tends to follow the same pattern. Thinking on that, the Data Enricher must be well developed mainly because it needs to provide a unique interface for any Data Provider API, making it easy to extend and not demand too much operational cost at each new integration.

Also, under the Data Enricher umbrella, we also need to define a good balance for the number of parallel requests performed to the external Data Source API that will allow us to read more data in a low-latency basis but at the same time won't compromise our Data Provider partners performance/availability.

Data Writer

When writing the data, we want to make sure we control the partitions of the When writing the data, we want to make sure we control the partitions of the data (see in Storage Structure). Not only that, but we also want to control the Data Enrichment Transactions that are still in progress. To keep track of the progress of the data, the Data writer must hold a counter that will be incremented each time it writes data to the output. A mechanism can be built on top of it to make sure the data consumer will only see the data when it's fully processed, see more on the Incomplete Enrichment Transactions section.

In the Data Writer process, we need to evaluate the rate at which we are going to output data, taking into account that the more frequently the data is written, more closer to real-time it is, but also more granular files it generates increasing the I/O time on the consumer side. It's a trade-off.

Serving Layer

In the final layer of the DES, there are the output data storage (AWS S3 bucket) and the query engine to access the data (AWS Athena). Also, since in the service input, it was offered two methods for ingesting data, it will be also offered the same methods so we can cover most company transactional services capabilities.

Storage strucure

The processing layer should write data in a structure that will allow them to be read by the consumers in a performatic and consistent way. Therefore, it was chosen the following partitioning schema for the data:

We want to make sure that whenever a query is executed, it won't scan the full table, so the 1st partition defined for the enriched_data table is the request_date (date when the data enrichment request was performed). As long as the consumer includes the date range in the query, the engine will only scan the related partitions.

Since we also want to group each Data Enrichment Transaction in a single data entity, the 2nd partition defined for the table is the erid field. With these two partitions defined, the executed queries will be able to retrieve the exact enriched data they will look for.

Incomplete Enrichment Transactions

The Data Writer phase is accountable for making sure the consumers will only see the data when it's fully processed and it can be done at the partition level. Essentially, while the row counter on the Data Writer is lower than number_of_records (parameter passed from the ingesting phase up to the processing job) the data will be saved in a partition with a suffix incomplete at the erid partition name. Whenever the records counter has reached the number_of_records from the request, it removes the suffix from the partition. Below is an example:

Serving data in a Push-oriented flow

For the services that have push capabilities, it was offered also a push-oriented approach for retrieving the data, build on top of AWS API Gateway and AWS Lambda. This component will be in charge of transforming HTTP requests in Athena queries executed over the output S3 Bucket data.

Each request will be composed by a date range (start_date, end_date parameters) and the erid field. For example, below is the HTTP request and the formed query:

```
# HTTP request
GET https://des.internal.sayprimer.com/v2/enriched_data?start_date=2024-02-
08&end_date=2024-02-11&erid=3c3f4b80-f728-4d7c-8567-1bc01df5f0a1

# Athena query
SELECT
    *
FROM    enriched_data
WHERE
    request_date >= '2024-02-08' AND request_date <= '2024-02-11'
    AND erid = '3c3f4b80-f728-4d7c-8567-1bc01df5f0a1'</pre>
```

This reponse will be composed by the payload (input data + enriched fields) as well as the fields to control the pagination for large datasets (number of pages and the next page, number of items per page, etc). Here is an example of a response data:

```
{
    "erid": "3c3f4b80-f728-4d7c-8567-1bc01df5f0a1",
    "data": [
        {
            "first_name": "John",
            "last_name": "Doe",
            "company_name": "sayprimer.com",
            "professional_email": "john.doe@sayprimer.com",
            "phone_number": "123456789"
        },
        {
            "first_name": "Foo",
            "last_name": "Bar",
            "company_name": "sayprimer.com",
            "professional email": "foo.bar@sayprimer.com",
            "phone number": ""
        },
        . . .
    ],
    "number_of_pages": 200,
    "next_page": 2,
    "status": "complete"
}
```

Status request and status check

The field status will play an important role in the Response, it has three possible values essentially:

- "inexistent": this status represents the absence of data. It might happen for two reasons: (1) a company service submits a data enrichment request to DES, but when requesting a GET on the same erid on the output API, the streaming process wasn't ready yet (resource allocation, temporary outage, etc) or (2) the requested erid don't exist, meaning the data enrichment request was never performed.
- "incomplete": this status represents an in-progress data enrichment transaction
- "completed": this status represents a completed data enrichment transaction

In the query of the previous section, only a single query statement was executed because it was enough to return more than zero rows, marked then as completed, however in a case where it returns zero records, the component must look at the incomplete transaction partitions (with _incomplete suffix) to know if the status of a transaction is either incomplete or inexistent. Here is an example of the same request, but now, zero records were found in the first query:

```
# HTTP request
GET https://des.internal.sayprimer.com/v2/enriched_data?start_date=2024-02-
08&end_date=2024-02-11&erid=3c3f4b80-f728-4d7c-8567-1bc01df5f0a1

# Athena query - zero records found
SELECT
   *
```

```
FROM
enriched_data
WHERE
request_date >= '2024-02-08' AND request_date <= '2024-02-11'
AND erid = '3c3f4b80-f728-4d7c-8567-1bc01df5f0a1'

# Athena query - more than zero records found
SELECT
*
FROM
enriched_data
WHERE
request_date >= '2024-02-08' AND request_date <= '2024-02-11'
AND erid = '3c3f4b80-f728-4d7c-8567-1bc01df5f0a1_incomplete'
```

In this particular case, the status will be considered incomplete. In case the 2nd query also doesnt't return rows, it will be marked as inexistent.

Since the Data Enrichment Transaction is asyncronous, the service that push the request, may also query the output endpoint from time to time in order to know whether the data is already available. Using the status mentioned above, the consumer may want to implement some rules in order to properly query the status of the request, so please consider the following python code:

```
# Data Enrichment Transaction request
post_data_enrichment_transaction(input_data, enrich_fields)
# ...
# Get Data Enriched request
response = get_enriched_data(start_date, end_date, erid)
attempt inexistent = 0
# Check for inexistent data
while(attempt inexistent < N MAX ATTEMPTS INEXISTENT and response.status ==
'inexistent'):
    wait_exponential_rate_seconds()
    response = get_enriched_data(start_date, end_date, erid)
    attempt_inexistent++
if(response.status == 'inexistent'):
    print("Data not found")
    # re-do the Data Enrichment Transaction request using another erid
else:
    # check for incomplete data
    attempt incomplete = 0
    while(attempt_incomplete < N_MAX_ATTEMPTS_INCOMPLETE and response.status ==</pre>
'incomplete'):
        wait some seconds()
```

```
response = get_enriched_data(start_date, end_date, erid)
if(response.status == 'completed'):
    break

attempt_incomplete++

# do the logic with the retrieved data.
```

It's important to set values for N_MAX_ATTEMPTS_INEXISTENT and N_MAX_ATTEMPTS_INCOMPLETE, so we don't keep the company service requesting the DES APIs forever. For N_MAX_ATTEMPTS_INCOMPLETE, it's important to recognize that depending on the dataset, we might set this to a higher value.

Serving data in a Pull-oriented flow

Both Data Fetcher (see in the Pull-oriented ingestion section) and Data Loader components were designed to be part of an ETL framework built to read/write data from/to many source systems. The orchestrator tool, AWS MWAA, is in charge of mediating the communication between the two jobs, essentially passing the erid from Data Fetcher to Data Loader.

With the erid, the job can start retrieving the data using Athena and writing data on the destination. Since it's built on top of an ETL framework with generic interfaces for both sources and sinks, the Data Loader will interact with the interfaces thal will talk to the inherited-specialized objects that will interact with the destination source. This will make the whole system more extensible and whenever new sources/destinations are added, it's only a matter of extending the interfaces and implement needed methods.

Config-based ETL framework

Whenever is possible, we want to avoid write customized code, making the service generic as necessary to attend many different use cases. In the case of pull-based ingestion, this isn't different; the engineering teams don't want to spend time writing new jobs whenever a new company service requests an integration for data enrichment. Based on this statement, we can create config-based workflows on the ETL framework on which each company microservice team should be able to add their own config files and the DES should be able to deploy automatically a new ETL pipeline for it. A config file can be similar to the following:

```
pipelineName: enrich_marketing_data
dataSource:
    type: mysql
    extraOptions:
        hostname: marketing-service.internal
        port: 3306
        password: ENV_PASSWORD
        ...
    table: users
    fields: [first_name, last_name, company]
    fields_to_enrich: [company_email, phone_number]
destination:
    type: mysql
    extraOptions:
        hostname: output-service.internal
```

```
port: 3306
password: ENV_OUT_PASSWORD
table: enriched_users
```