Airflow Implementation Documentation

Version 1.0

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# How to enable the user to run the dag with his own connection name

## Problem:

User needs to be able to run the dag with his own credentials or in airflow terms, a different connection names.

## Solution:

Firstly, we need to keep in mind that airflow is not talking to the underlying jobs ‘directly’. There is an API layer that sits between the airflow task and the job. Airflow hits this REST API and the API in turn triggers the job.

Having said that, it shouldn’t matter to the API that which user airflow is using to hit the API. What will matter is that what API credentials the Airflow connection is using to initiate the connection and make the post request.

Consider the example below:

The value that is provided in the ‘http\_conn\_id’ option of the task specifies the connection that this task will use to initiate the connection and make the post request to the API layer. This connection name is specified in the Airflow’s Connections sections and all the authentication information is provided in there. The user can just provide his own connection name in place of ‘pdill\_datamasking’ and the task is good to use for that connection value.

datamasking\_ingestion = SimpleHttpOperator(

    task\_id='data\_ingestion',

    http\_conn\_id=pdill\_datamasking,

    endpoint="/datamasking/ingestion/{{ task\_instance.xcom\_pull(task\_ids='get\_configurationId', key='return\_value') }}",

    method='GET',

    response\_check=lambda response: response.json()['status']=="SUCCESS",

    log\_response=True,

    dag=dag

)

# Airflow Code Demo

Basic Idea : The dag is a sequence of tasks that have dependencies between on another. In the pdill dag we have http operators that trigger the jobs and then we have sensors to track the progress of the job. These sensors halt the execution of the dag until the underlying job has completed and only proceeds if the job results in a success

from datetime import datetime

from airflow import DAG

import json

from airflow.operators.http\_operator import SimpleHttpOperator

from airflow.operators.python\_operator import PythonOperator

from airflow.sensors.http\_sensor import HttpSensor

Firstly, We import the required libraries such as the DAG class and other operators which are required for the development.

We can also import custom libraries from custom packages here.

Summary - Importing libraries and operators and custom functions

default\_args = {

'description':'TDMF data masking',

'start\_date': datetime(2020,9,25),

'email': ['prashanth.desani@wellsfargo.com'],

'retries': 3,

'email\_on\_failure': True,

'email\_on\_retry': False,

}

Summary: Default arguments are required and are sent to all tasks. We can set options that are redundant in all tasks here

The default arguments are the arguments or options that will be passed to all the tasks in the Dag. This will help us avoid redundantly providing these options to each and every task separately. It is also useful if we need to make small changes in these options from time to time. We will have to make it in just one place and not having to make changes in all the tasks.

Description: This parameter holds a one-line description that will show up on the UI beside your dag name (which is defined later in the code).

Start\_date: start date is defined for the scheduler to understand from which day it is supposed to start executing this particular dag. It is helpful when you want to start the dag on a certain day post-deployment or otherwise you can set any past date. However, this is an important option. There is another option catchup (defined later in the DAG object) which goes in conjunction with this start date. The catchup option tells the scheduler if it is supposed to ‘catch-up’ on executing the historical dag runs or not. So if you set a historical date say 5 days ago from the date of deployment. The scheduler will create 5 dag runs and the next required run when the scheduled interval arrives

Email: The email option defines a list of emails that the task is supposed to send the emails to.

Retries: The number of times the task should retry before giving out a failure status

Email\_on\_failure: This option tells the scheduler if it is supposed to email on the failure of a task

Email\_on\_retry: each task also needs retry options. The task then retries for the specified number of retries and this option just tells the scheduler if it is supposed to send an email to the email list or not.

configuration\_id = ""

pdill\_datamasking\_bkp='pdill\_datamasking\_bkp'

These are initialization variables just created for passing them to the tasks.

def resolve\_config\_id(\*\*kwargs):

task\_instance = kwargs['task\_instance']

configuration\_dict = json.loads(task\_instance.xcom\_pull(task\_ids='data\_extraction', key='return\_value'))

configuration\_id = configuration\_dict['configurationId']

return configuration\_id

Summary - This is a Custom Function that helps us resolve the configuration ID. You can define any number of functions and call them through the Python Operator

The resolve\_config\_id function resolves the configuration\_id for the consequent steps in the dag. It pulls the return value from the starting task and then pushes it as a singleton value in the Xcom to make it easily accessible to the consequent tasks.

Xcom is a common Dictionary available through all the tasks in the dag. This helps the tasks to talk to each other and pass variables and make decisions on the outcome of the previous

Note: the tasks of a Dag are rendered every time a dag run is created. Hence some parameters which are dynamic need to be passed in the ‘Jinja Templating’ syntax. Jinja Template Engine is a web page rendering engine that resolves native python variables to our tasks. You can find them enclosed in ‘{{ … }}’ these braces. It is also important when you want to use variables that are passed on runtime. These runtime variables are attached to the dag\_run object which is only available during runtime.

def monitor\_job(response):

print('###################################monitor process response ##################')

print(response.text)

responseJson=json.loads(response.text)

jobStatus=responseJson["status"]

if jobStatus == "FAILED":

raise AirflowException("Job FAILED.")

else:

return jobStatus=="SUCCESS"

Summary : This is also a custom function that is invoked by the HttpSensor. It processes the response received and gives either True or False based on condition.

This function is defined to be called in the response check part of the HttpSensors (defined later). It takes in a response dictionary and then prints out the entire message in the log giving us a better idea of all the things the response contained.

dag = DAG('pdill\_datamasking', schedule\_interval=None, catchup=False, default\_args=default\_args)

Summary : The dag object is created here with the required parameters.

This line creates the dag object to which all the tasks will be attached and execute.

The first argument is the name of the dag (which will show up on the UI),

next is the scheduled interval which defines the schedule of the dag in the ‘cron’ format

Next is the catchup option to tell the scheduler if historical dag runs are to be created or not

Lastly, we have the default\_arguments dictionary wherein we pass the default arguments that are needed in all the tasks and we do not want to keep passing them in each one separately

post\_data = {

"app": "CNAPP",

"project": "SS",

"sor": "msp",

"datasetName": "CHARMING\_17",

"date": "2021-05-30"

}

This is a data dictionary sent to the POST endpoint at the start of the execution. It is hardcoded as of now but will be parameterized

datamasking\_pipeline\_execution=SimpleHttpOperator(

task\_id='datamasking\_pipeline\_execution',

http\_conn\_id=pdill\_datamasking\_bkp,

endpoint='/datamasking/execution',

headers={"Content-Type": "application/json"},

data=json.dumps(post\_data),

response\_check=lambda response: response.json()['configurationId'] != None,

xcom\_push=True,

log\_response=True,

dag=dag

)

Summary: This the the SimpleHttpOperator, that helps us send Post and get requests to the api

Now creating a task in Airlfow means just invoking the right operator object you want with the right options. Here we invoke the SimpleHttpOperator which we imported at the beginning and provide the right options to it.

The task\_id defines the name of the task. It should be unique throughout the DAG.

The next is the http\_conn\_id. This is the connection id that we define in the ‘Connections’ section of the DAG UI. This helps us decouple secret information from the DAG’s codebase.

The next is ‘endpoint’ which is the endpoint we’re trying to hit. Note that it does not have the hostname attached to it.

The headers are the headers to be passed with your request.

Data indicates any data you want to pass with the request. In our case, it is the post\_data dictionary that we have defined earlier.

Moving on we have the response\_check option it checks for the required success criteria and only gives a success output if it is met otherwise the task returns a failed state.

Xcom\_push value indicates if the return\_value is supposed to be pushed to the xcom or not.

Log\_response value indicates if the task is supposed to log the entire response or not.

Lastly, the dag option indicates which dag this task belongs to

get\_configurationId = PythonOperator(

task\_id='get\_configurationId',

dag=dag,

python\_callable=resolve\_config\_id,

provide\_context=True

)

Summary: this is a python operator that resolves the configuration ID and pushes then to Xcom.

This is a python operator that helps us execute custom python functions. Here we are using it to invoke resolve\_config\_id function that we have defined.

datamasking\_extraction = SimpleHttpOperator(

task\_id='datamasking\_extraction',

http\_conn\_id=pdill\_datamasking\_bkp,

endpoint="/datamasking/extraction/{{ task\_instance.xcom\_pull(task\_ids='get\_configurationId', key='return\_value') }}",

method='GET',

response\_check=lambda response: response.json()['status']=="SUCCESS",

log\_response=True,

dag=dag

)

Summary: this is again a Simple Http Operator that fires a get request to the datamasking/extraction endpoint.

Note : Response check in Simple Http Operator pertains to that request only. This means the operator only triggers the job and then exits with success when the response\_check condition is satisfied. This is why we use a sensor to track progress

This is similar to the previous SimpleHttpOperator but with a few differences.

The first one is that we attach a configuration\_id dynamically based on what we receive from the spark UI.

The next is the ‘method’, we specify it as ‘GET’. If we do not specify this value it defaults to ‘POST’.

datamasking\_extraction\_jobstatus = HttpSensor(

task\_id='datamasking\_extraction\_jobstatus',

http\_conn\_id=pdill\_datamasking\_bkp,

endpoint="/datamasking/jobstatus/{{ task\_instance.xcom\_pull(task\_ids='get\_configurationId', key='return\_value') }}",

method='GET',

request\_params={},

response\_check=lambda body:monitor\_job(body),

#response\_check=lambda response: response.json()['status']=="SUCCESS",

poke\_interval=5,

dag=dag

)

Summary : the http sensor pokes the tracking endpoint after the set interval of 5 seconds. This interval can be changed based on our requirements.

This is a HttpSensor. We have imported this at the start and it is used to keep poking any endpoint until we get the required response from the server. So here we have an endpoint that pokes with the interval of 5 seconds and it will keep poking it until we get the ‘status’ key in the response as ‘SUCCESS’

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datamasking\_pipeline\_execution >> get\_configurationId >> datamasking\_extraction >> datamasking\_extraction\_jobstatus >> datamasking\_execution >> datamasking\_execution\_jobstatus >> datamasking\_validation >> datamasking\_validation\_jobstatus >> datamasking\_ndm >> datamasking\_ndm\_jobstatus >> datamasking\_ingestion >> datamasking\_ingestion\_jobstatus >> datamasking\_complete

Summary : The above is the hierarchy that we have defined. The hierarchy can be any way we want. It can have branches and also conjunctions. In our case, we did not need any of them so we have a pretty simple hierarchy. The ‘>>’ operator defines a dependency between the tasks.

# Email Code Explanation

The email code is encapsulated in a function called as a notify email. This function is called in the on\_failure\_callback and on\_success\_callback in the default arguments of the dag (This means that these options are going to be passed to all the tasks in the dag). Thus all the tasks will call the ‘notify\_email’ function when they succeed or when they fail.

def notify\_email(kwargs):  
 *"""Send custom email alerts."""*

ti = kwargs['ti']  
 dag\_run = kwargs['dag\_run']  
 var = kwargs['var']['json']  
 params = kwargs['params']

The above variables are available in the kwargs which is like a global dictionary available for us to get metadata about the dag and the tasks.

Here in the above snippet, we have used the following.

“ti” : ti stands for task\_instance that is the instance of the task from where the function is called.

“dag\_run” : dag\_run is a global object that holds information about the execution\_date the run\_id and other dag related information.

“var” : This variable is the configuration that we pass to the dag. Like the SOR name and everything. Basically everything that we give while triggering the dag. Or the parameters of the

dag basically.

recipient\_emails = ['prashanth.desani@wellsfargo.com']

The above is the variable where we store the recipient emails. The people airflow needs to send emails to whether on success or failure.  
  
 logs\_link = '{}/log?task\_id={}&dag\_id={}&execution\_date={}'.format(dag\_run.conf.get("webserver", "base\_url"),  
 ti.task\_id, ti.dag\_id, ti.execution\_date)

The above creates a log link. When anyone receives this link and clicks on it. They will be redirected to the logs page for that task on the UI directly from the email.

title = ''  
 body = ''  
  
 # email title.  
 if dag\_run.\_state == "success":  
 title = f"[ Airflow Success Alert ]: {dag\_run.dag\_id} Execution Completed"  
 body = """  
 Hi Everyone, <br>  
 <br>  
 Job {dag\_id} has executed successfully.<br>  
 <br>  
 Thank you,<br>  
 Airflow Bot <br>  
 """.format(dag\_id=dag\_run.dag\_id)  
 else:  
 if ti.state == "failed":  
 title = f"[ Airflow Failure Alert ]: {ti.task\_id} Failed for {dag\_run.dag\_id} failed"  
 body = """  
 Hi Everyone, <br>  
 <br>  
 Task {task\_id} failed.<br>  
 Please check the log at : {log\_link}<br>  
 <br>  
 Thank you,<br>  
 Airflow bot <br>  
 """.format(task\_id=ti.task\_id, log\_link=ti.log\_url)  
 elif ti.state == "success":  
 title = f"[ Airflow Success Alert ]: {ti.task\_id} Succeeded for {dag\_run.dag\_id}"  
 body = """  
 Hi Everyone, <br>  
 <br>  
 Task {task\_id} has completed successfully<br>  
 <br>  
 Thank you,<br>  
 Airflow bot <br>  
 """.format(task\_id=ti.task\_id, log\_link=ti.log\_url)  
 else:  
 raise AirflowException('{} task state is not supported in email notifications'.format(ti.state))

The above is a simple snippet that handles the the creation of the email template using multiple variables. Each object has a different metadata about the dag and the tasks.

send\_email(recipient\_emails, title, body)

Once the above template is created the email is simply sent using the send\_email function.

# Operator Factory Code

# Where are the logs stored?

The logs are stored in the form of text files on the server where airflow is running. It has a specific folder structure while storing the logs. The logs are stored in the following way.

Logs

Dagname

Taskname

<retrynumber>.txt

Although the same is available in the UI in the logs section of the task.

How can I run the job from the point of failure?

On the UI, you must navigate to the DAG that you want to re-run. Once you’re on go to the graph view and there you can see the tasks highlighted in red as the failed tasks. To rerun the task you need to click on the task and a window pops up with a lot of options. One of them is a clear option. It also has sub-options. They are there for 2 cases. Case 1 is that you just need to rerun that one task. In this case uncheck the ‘Downstream’ and ‘Recursive’ suboptions. The other case which is most likely is that you want to clear/rerun the failed task and also all the downstream tasks. In this case remember to check the ‘Downstream’ and ‘Recursive’ options.

Performance Improvement?

Airflow’s performance largely depends on what mode it is running on and what backend it is using. Airflow has metadata tables that it uses to make execution decisions

# DAG Triggers and Configuration Files

## Possible DAG trigger Scenarios

### A single one-time dag run for each configuration

* In this scenario one dag will run for each configuration that is generated by the EDL console.
* Multiple instances of the same dag will run in parallel when there are multiple configurations
* Each configuration progress is handled by a separate dag run instance and progress of the configuration file can be tracked through the progress of the dag run that is triggered for that specific configuration id

### A single one-time dag run on a schedule to run with the same configuration

* In the case where the dag is supposed to be run on a schedule with the same configuration, a metadata table can store the schedule and the configuration id
* A separate dag apart from the PDILL dag will run continuously, checking this metadata table to trigger the pdill dag for the respective configurations as and when their schedule arrives
* Alternatively, separate dags can be written for each configuration ID and then airflow’s scheduler can handle the triggering of those dags

### A single dag run in parallel for multiple configurations

* In this scenario, tasks can be dynamically generated based on the number of configuration\_ids we receive from the GET API
* Multiple branches of the tasks will be generated dynamically and each branch will process a particular configuration\_id

### Multiple dag runs for multiple configurations

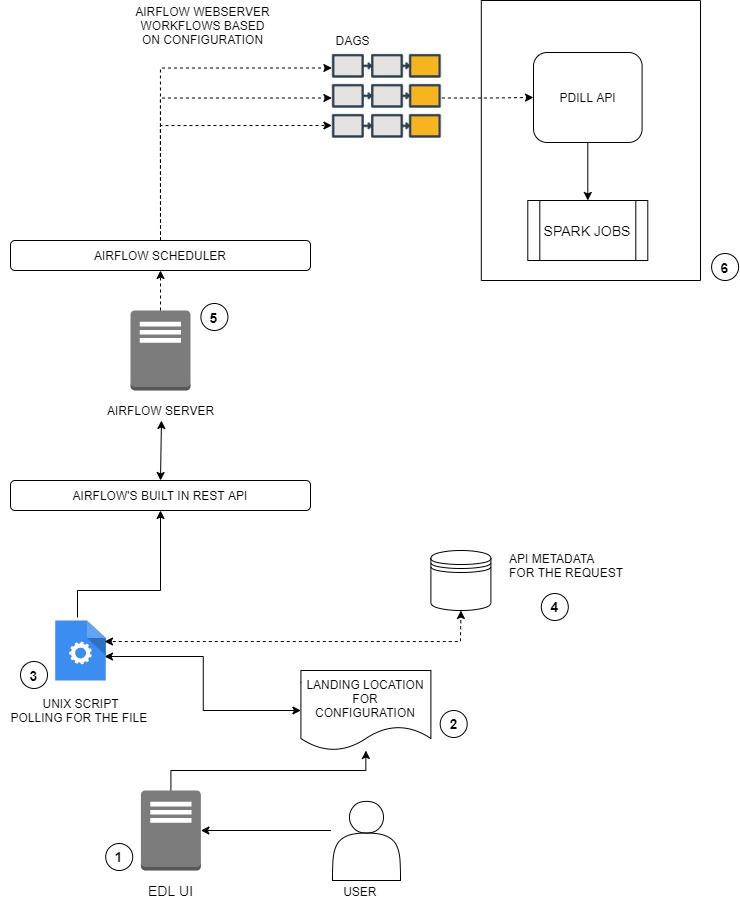
* This is the most simple, where we trigger multiple workflows for multiple configurations
* Each instance of the PDILL dag will process one configuration ID

### Getting multiple jobs from the GET method API

### Trigger Mechanism

#### Unix Script Running on the server

A unix script running on the server where the configuration file. This script will poll for the configuration files or payload and then trigger the PDILL Dag.



1. The user using the EDL UI creates a configuration file which is stored on a local storage.
2. The EDL console creates a landing location for the configuration file.
3. The Unix script will poll this landing location continuously for new configuration files and trigger the dags automatically based on a metadata or fields defined in the configuration file.
4. The meta data for the triggering of the dags can also be stored on a Database.
5. Airflow dags running on the airflow webserver are triggered using the Airflow REST API.
6. Airflow dags that are triggered send a POST request to PDILL API that processes the data using underlying jobs.

##### Questions

1. How many servers will airflow need to talk to in order to trigger workflows? *This question is regarding the possibility that Airflow might have to trigger dags on different servers. If yes, we would need to set up the connections to all of those servers in order to trigger* something on that server, just like we have set up for the API that we have now.
2. What would the configuration file contain?
3. Is the EDL configuration file a one-time use configuration?
4. Will the configurations be unique? *Multiple users can create configurations at the same time, right? What would happen in the event when there are duplicate configurations created?*
5. Is there a naming convention defined for the configuration file?
6. On what basis can I decide which dag is supposed to be triggered?

#### Separate PDILL DAG running to check for configuration files

This approach is better when we need to poll for configuration files on different unix servers. The dag will need to SSH into the servers and then poll for the files on those servers and then trigger the files on the airflow server

## Problem Statement:

Airflow is receiving multiple configuration items in response. Our Airflow DAG needs to be able to handle the multiple configuration items in the response that we are getting.

## Assumption:

* The configuration item is unique
* The configuration item is one-time use only
* Dag will be triggered based on the availability of the configuration item

## Solution:

Airflow operators can be dynamically generated based on the number of configuration items present in the response received by the API. The jobs will run in parallel for each of the configuration items and the branches will be created automatically

Another solution to this problem can be that our DAG can write other DAGs. Consider this as a master DAG and this DAG will write standalone DAGs for us based on the configurations received. This solution is only valid if we have to schedule some configuration items because they will have to run on a schedule and the same dag cannot run on different schedules.

# PDIL DAG Design

The screenshot below describes the graphical view of the Airflow Dag for PDILL

<insert DAG screenshot>

|  |  |  |
| --- | --- | --- |
| Operator Name | Function | Parameter Details |
| data\_extraction | This operator hits the endpoint for extraction of data and generates CSV files | *task\_id:* Name of the task that is visible on the DAG and for Xcom push and pulls.  *http\_conn\_id:* Name of the connection id as defined in the ‘Connections’ section of Airflow UI with auth details  *endpoint:* The endpoint that the operator will hit. The endpoint does not include the server name in it only the name of the route is mentioned. Airflow picks up the server name from the Connections sections as mentioned in the Airflow UI  *headers:* The headers for the request  *data:* The payload for the request in a valid JSON string  *response\_check:* This parameter checks the value of a particular parameter in the response to make a decision if the request was completed successfully or failed  *xcom\_push:* pushes to values to the Xcom  *log\_response:* The response of the request gets logged in the logs section of the UI  *dag:* The dag where the operator belongs |
| get\_configurationId | This operator is responsible for resolving the configuration id required to be appended to the ‘GET’ requests in the subsequent tasks in the DAG.  This operator just receives the configurationId from the Xcom pushed by the *data\_extraction* operator which is in dictionary format and pushes it to Xcom as a singleton value for it to be easily accessible by the subsequent operators | *task\_id:* Name of the task that is visible on the DAG and for Xcom push and pulls.  *dag:* The dag where the operator belongs  *python\_callable:* The python function to be called for the execution |
| data\_masking | This operator starts the spark job responsible for data masking.  The configurationId is received from the Xcom and appended to the endpoint making it dynamic for any configuration Id | *task\_id:* Name of the task that is visible on the DAG and for Xcom push and pulls.  *http\_conn\_id:* Name of the connection id as defined in the ‘Connections’ section of Airflow UI with auth details  *endpoint:* The endpoint that the operator will hit. The endpoint does not include the server name in it only the name of the route is mentioned. Airflow picks up the server name from the Connections sections as mentioned in the Airflow UI  *method:* ‘POST’ or ‘GET’ (or other HTTP verbs)  *headers:* The headers for the request  *data:* The payload for the request in a valid JSON string  *response\_check:* This parameter checks the value of a particular parameter in the response to make a decision if the request was completed successfully or failed  *xcom\_push:* pushes to values to the Xcom  *log\_response:* The response of the request gets logged in the logs section of the UI  *dag:* The dag where the operator belongs |

# Airflow Trigger Mechanism

## Solution Overview

## Questions:

# Outstanding Questions:

How does the DAG trigger from the UI?

How does one run a DAG multiple times sequentially?

How to control DAG Runs?

How to Promote code from dev to test environment and then finally to production?

How is unit testing done?

# User Stories

|  |  |  |  |
| --- | --- | --- | --- |
| User Story | Task | Acceptance Criteria | Description |
| Install Airflow Using Rabbit MQ and Redis with Celery | Configure Postgres/MySQL for airflow backend | Postgres/MySQL server is setup and running |  |
| Setup the machine with Ubuntu | Machine is up and running and ready to install airflow |  |
| Install Airflow | Airflow Webserver and Scheduler is running successfully |  |
| Code Optimization | Include comments and remove redundant functions | Comments provide a complete insight of the code functionality |  |
| Create Operator Factory | Dag Operators are rendered and generated in multiple Dags as per requirement | This allows Code reusability for airflow dags |
| Code Parameterization | Parameters can be passed to the DAGs along with the trigger to run DAGs for different configurations |  |
| Code Review and Code Promotion | Create a repository for code | Repository setup with access to create branches based on the main/master branch |  |
|  | Code review sessions for already build DAGs | PR raised, approved for the code and merged to main/master branch |  |

# Java Code

## AirflowController.java

### Overview

This file contains all the utilities and endpoints required to control the underlying Talend/Delphix Jobs. We have several endpoints available. Most of the endpoints are the ones that we are already using in Airflow DAG. The one’s we have not used in our DAG are:

## Endpoint: /datamasking/joblist/{app}/{jobDate}

Does the same work as listjobs. Just the name of the endpoint is different. Looks like a duplicate endpoint. Let me know your thoughts.

## Endpoint: /datamasking/listjobs/{app}/{jobDate}

This endpoint takes the ‘app’ name and the ‘jobDate’ from the request as parameters.

// A new array list is created named jobs.

List<Configuration> jobs = new ArrayList<Configuration>();

// The dao.scan() gives us all the requests available.

        List<String> allRequests = dao.scan();

// Below line just prints out the number of requests.

        log.info("Data masking configurations scanned {}", allRequests.size());

// filter out deleted ones for the ‘app’ that is provided in the get request and the project name as ‘AF’

        List<Configuration> ll = allRequests.stream()

                .map(this::convertConfiguration)

                .filter(Objects::nonNull)

                .filter(c -> !JobStatus.DELETED.equals(c.getStatus()))

                .filter(c -> app.equalsIgnoreCase(c.getApp()) && "AF".equals(c.getProject()))

                .collect(Collectors.toList());

// A new Configuration is created and then added to the arraylist of ‘jobs’ which is later returned as the response to the request.

        for (Configuration \_conf : ll) {

            Configuration conf = new Configuration();

            conf.setConfigurationId(\_conf.getConfigurationId());

            conf.setProject(\_conf.getProject());

            conf.setApp(\_conf.getApp());

            conf.setSor(\_conf.getSor());

            conf.setDatasetName(\_conf.getDatasetName());

            jobs.add(conf);

        }

// The jobs are returned to in the response.

        return jobs;

### Questions:

1. What are the different types of the jobs that are returned to us in the list?

2. Does it give us all the jobs except ‘DELETED’ jobs for that particular date?

### Comments:

We need to run this endpoint in Postman and then find out what list of jobs we get in the response. Based on this, we can infer if need additional filtering while selecting which job to trigger.

## Endpoint: /datamasking/startJobs

I did not quite understand what this does. Initial analysis is that it starts some masking of the data using Delphix. It needs some csv attached with the request.

## Endpoint: /datamasking/submit

This endpoint submits the jobs for masking. It prepares the masking jobs for execution.

## Endpoint: /datamasking/executes

This endpoint scans all the configurations and then prepares and submits the jobs for execution.

# AirflowTaskController.java

Same as AirflowController.java

# DataMaskingConfigurations.java

This file deals with the endpoints related to configurations creations and user access.

The code in the preAction takes into account from which user the request is coming in. Hence here we will have to create a universal user, or we would need separate connections for each user that need to trigger masking jobs.

Recommended Approach: have a universal user that only Airflow can use and that can load all the tables in the lower environment.

protected void preAction(HttpServletRequest request, HttpServletResponse response) {

        log.info(

                "Request From: " + request.getRemoteUser() + " / userId " + userUtil.getUserId(request) + " Access: " + request.getRequestURI() + " "

                        + request.getQueryString());

        log.info("Request IP: " + request.getRemoteAddr() + " Request sessionId: " + request

                .getRequestedSessionId());

    }

# DataMaskingDao.java

It is a data access object file for Data Masking essential objects. Dao pattern is a structural pattern that allows us to isolate the application/business layer from the persistence layer (usually a relational database but could be any other persistence mechanism) using an abstract API.

# DataMaskingDbDao.java

This file has the CRUD for the database functions such as load extract and other features that you would need.

# TalendRequestConfiguration.java

# Architecture Findings

We have clarity to some extent on how Airflow can fit in to this architecture. We have endpoints that can give us a list of jobs that are yet to be executed. Using this list of jobs we can trigger the PDILL dags for each configurtaions/application.

Although User access is still somewhere, we do not know much about. There is part of the code where the code checks which user is sending this request. However as we rightly suspected beefore the user that is referenced in the code is the user from the request and not the unix user. So we have to either have a single universal Airflow connection id user or multiple connection ids setup for each of the user and then pass this ID as a parameter.

# Todo Items

Need to mention the commands and how the code is managed

* How the code is going to be promoted -- done
* How the code is going to be promoted through Urban deploy -- work with Prashanth

Celery Executor - How workers will be assigned on multiple edge nodes.

Need to talk about RabbitMQ and Redis and recommendation

* How to start
* How to maintain
* How to work with Celery
* Send new JIRA stories with description for next week’s work - IMPORTANT Wednesday
* Put details in the document
* Code Optimization
* Multiple Dag run poc
* PPT - P4