

```
FROM AIRFLOW IMPORT DAG
```

Airflow the perfect match in our Analytics Pipeline

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Senior Business Intelligence Architect @LOVOO

Airflow 
Summit 2020

AGENDA

1. Why we met?
2. How we met?
3. The first date!
4. Fun dates!
5. Is there any dynamic in between?
6. Recap and conclusion

About LOVOO

- LOVOO is a dating and social app and the place for chatting, live streaming, watching streams and getting to know people.
- Germany - Dresden & Berlin - 2011
- Acquired by The Meet Group (NASDAQ:MEET) in 2017
- Top 3 Dating App in Europe
- + 280 TB of Data
- ~ 6 TB Monthly Growth
- + 3 TB daily total aggregated data
- + 36 TB Swipes (162,824,303,474)

Analytics

- 1 Head
- 6 Data Analysts
- 2 BI Architects



- Product
- Finance
- Marketing
- Talent Management
- Customer Insights
- CRM

What can you expect?

My main purpose today is to tell you about our journey with Airflow as well as a few different use cases that could also boost the work of your Analytics/BI team on a daily basis.

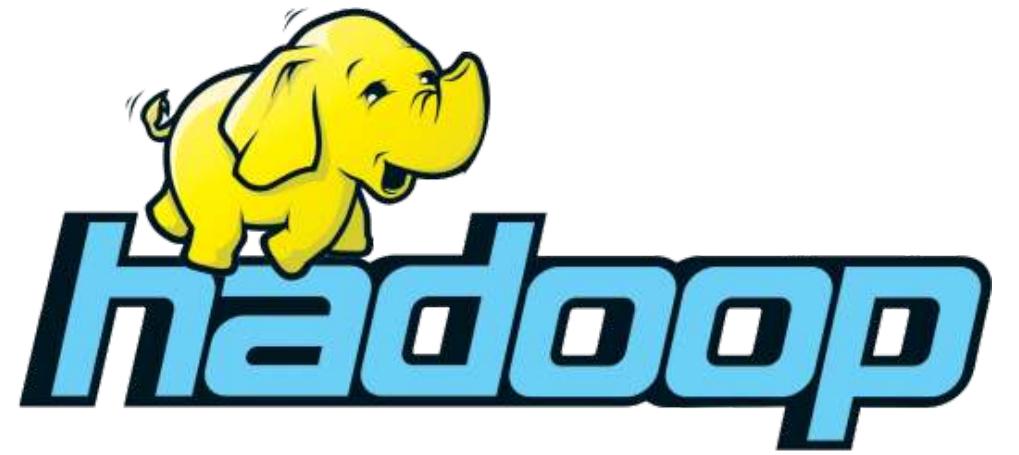
- Pieces of code (examples)
- Way too many screenshots

AGENDA

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On-premise

cloudera

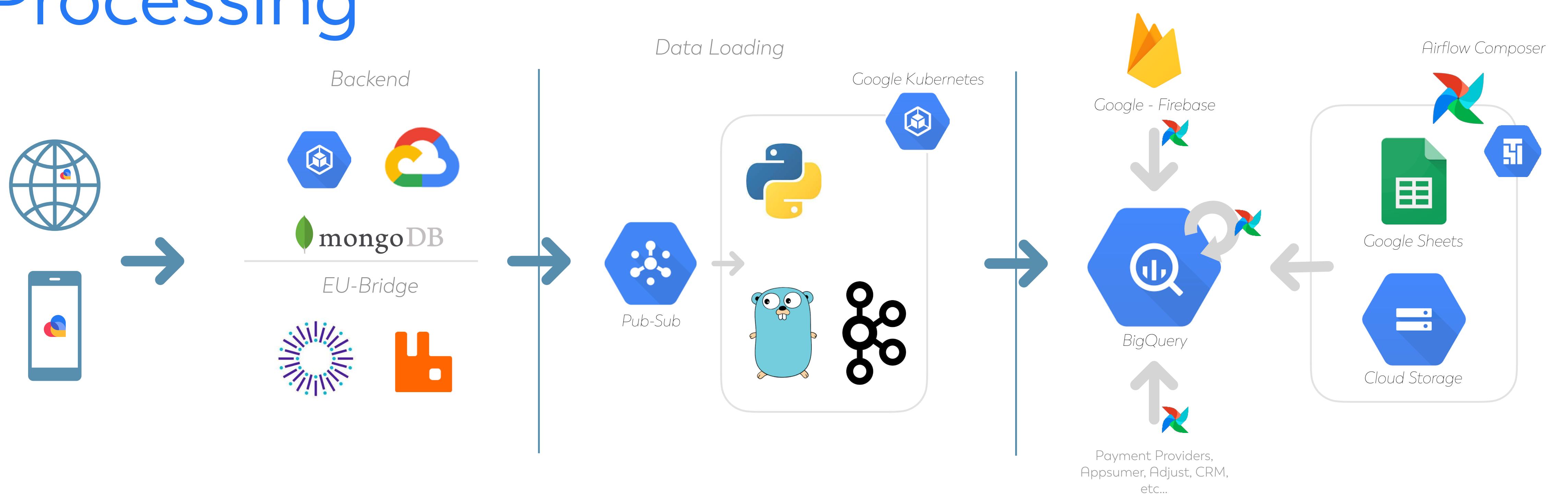


We went Cloud

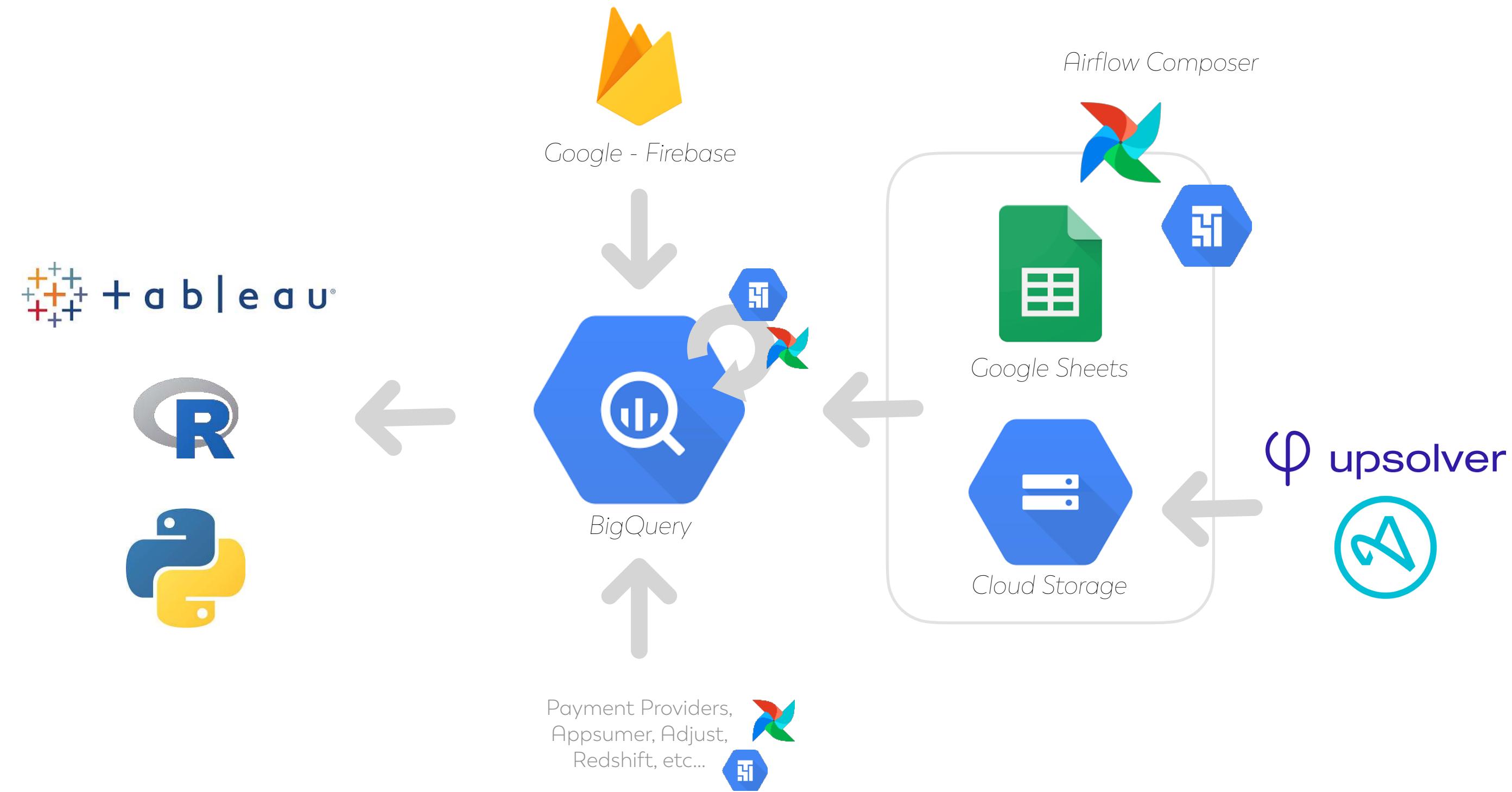


THE PROFILE DETAILS...

Data Processing



Analytics Data-Core



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Orchestration Tool

- Identify what is out there
- Costs?
- Scalability?
- Data sources compatibility?
- Knowledge/Human Resources?

RIGHT SWIPE...

Airflow



- Great community
- Game changer
- Mobile App
- Python
- BigQuery

Google Cloud Composer



- Fully Managed Airflow
- Scalable
- IAP - Secure
- Focus on building the Analytics data pipeline
- Ease of implementation

Google Cloud Composer



- Fully Managed Airflow



This is an alpha release of Cloud Composer. This product might be changed in backward-incompatible ways and is not recommended for production use. It is not subject to any SLA or deprecation policy. This product is not intended for real-time usage in critical applications.

- Focus on building the Analytics data pipeline
- Ease of implementation



⚠ Confidential Material: This page is confidential. Do not share or discuss until authorized to do so.

AGENDA

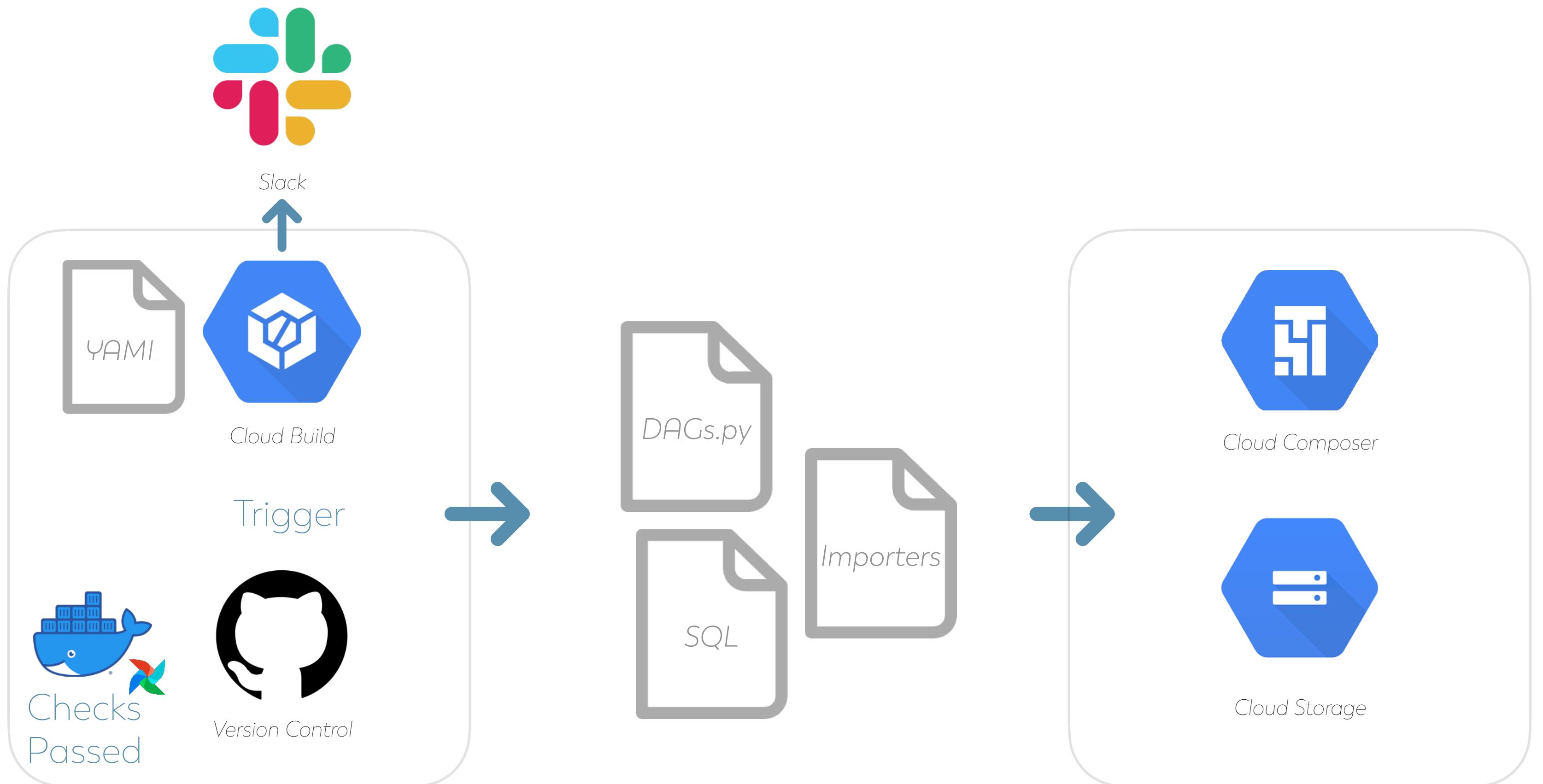
1. Why we met?
2. How we met?
- 3. The first date!**
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TODO List

- SQL Scripts → Data Modeling
- DAGs
- Permissions - Service Accounts
- Data Importers
- Create a Composer Environment
- How do we deploy? → CI/CD

GROWING TOGETHER!

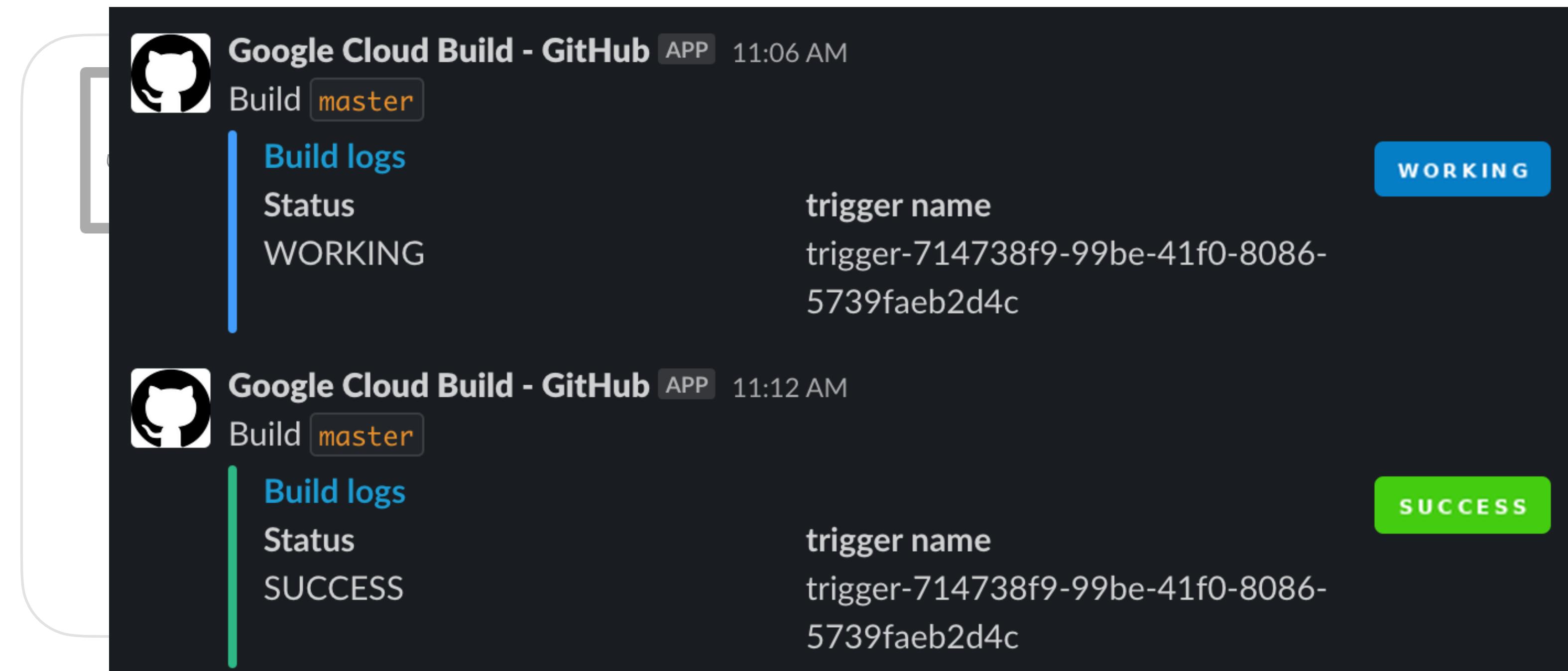
CI/CD



CI/CD



Slack



Google Cloud Build - GitHub APP 11:06 AM
Build master

Build logs
Status
WORKING

trigger name
trigger-714738f9-99be-41f0-8086-
5739faeb2d4c

WORKING

Google Cloud Build - GitHub APP 11:12 AM
Build master

Build logs
Status
SUCCESS

trigger name
trigger-714738f9-99be-41f0-8086-
5739faeb2d4c

SUCCESS

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HOW DOES IT LOOK LIKE?

DAGs

Operators



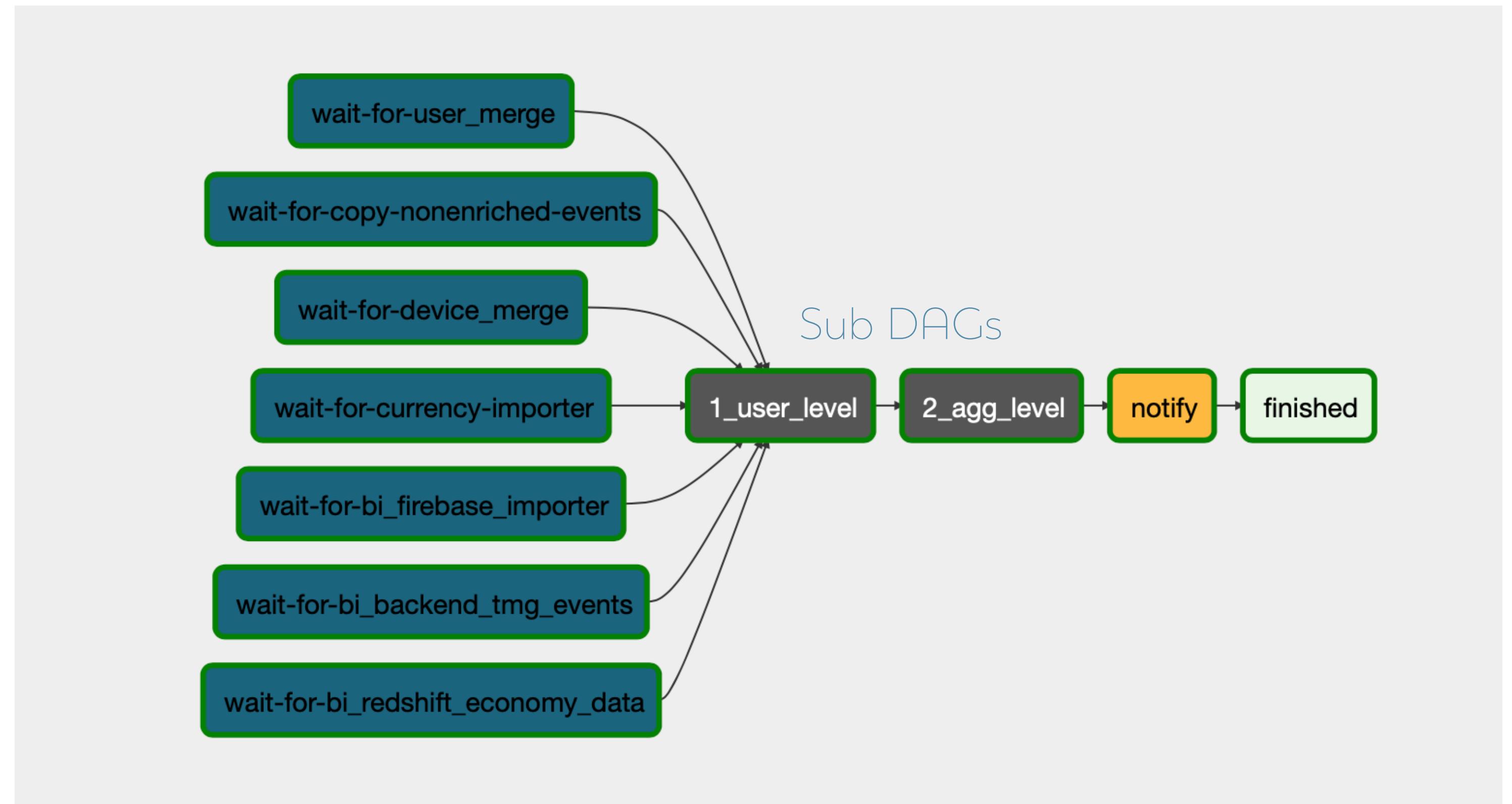
- 26 DAGs
- Sub-DAGs
- Branching
- Jinja Templating
- Hooks
- Pools
- Trigger rules

	DAG	Schedule
<input checked="" type="checkbox"/> On	adevents-repair	0 4 * * *
<input checked="" type="checkbox"/> On	airflow_monitoring	None
<input checked="" type="checkbox"/> On	analytics_jobs	0 5 * * *
<input checked="" type="checkbox"/> On	analytics_jobs_live	0 5 * * *
<input checked="" type="checkbox"/> On	antispam-creditfarm-detection	@daily
<input checked="" type="checkbox"/> On	antispam-reputation-modeltraining	@daily
<input checked="" type="checkbox"/> On	appsumer-importer	00 11 * * *
<input checked="" type="checkbox"/> On	appsumer-importer-hayi	00 12 * * *
<input checked="" type="checkbox"/> On	bi_backend_tmg_events	30 2 * * *
<input checked="" type="checkbox"/> On	bi_data_check	40 3 * * *
<input checked="" type="checkbox"/> On	bi_firebase_importer	40 4 * * *
<input checked="" type="checkbox"/> On	bi_firebase_live_events	40 4 * * *
<input checked="" type="checkbox"/> On	bi_marketing_events_jobs	50 6 * * *
<input checked="" type="checkbox"/> On	bi_marketing_jobs	50 9 * * *
<input checked="" type="checkbox"/> On	bi_payment_provider_apis	40 4 * * *
<input checked="" type="checkbox"/> On	bi_redshift_economy_data	30 4 * * *

PRETTY ON THE OUTSIDE...

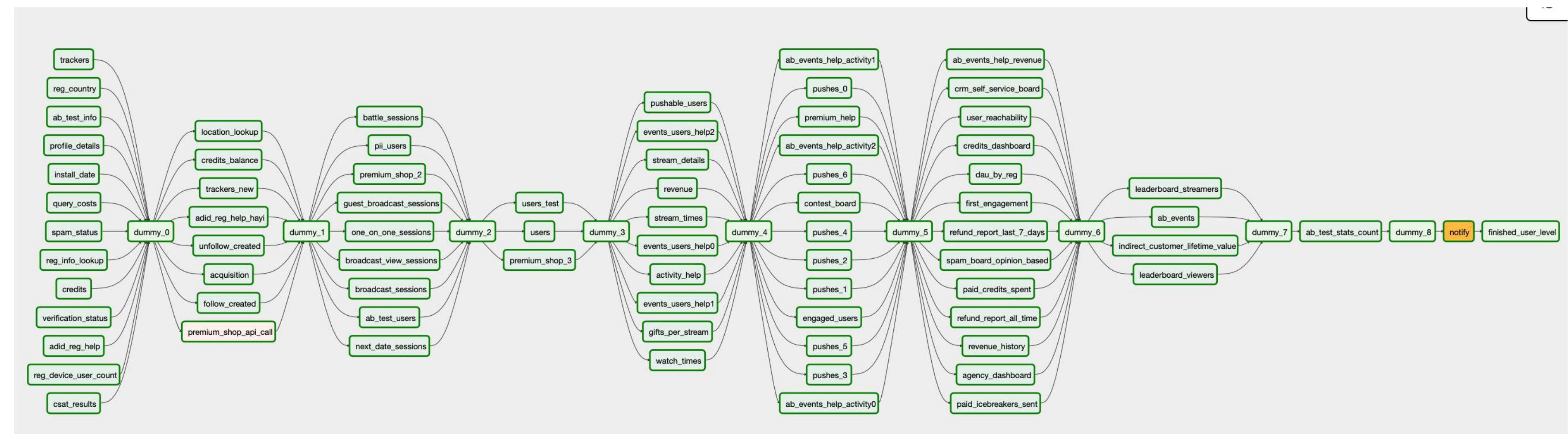
The Core

Analytics - Workflow



The Core

Sub DAG



Reports!

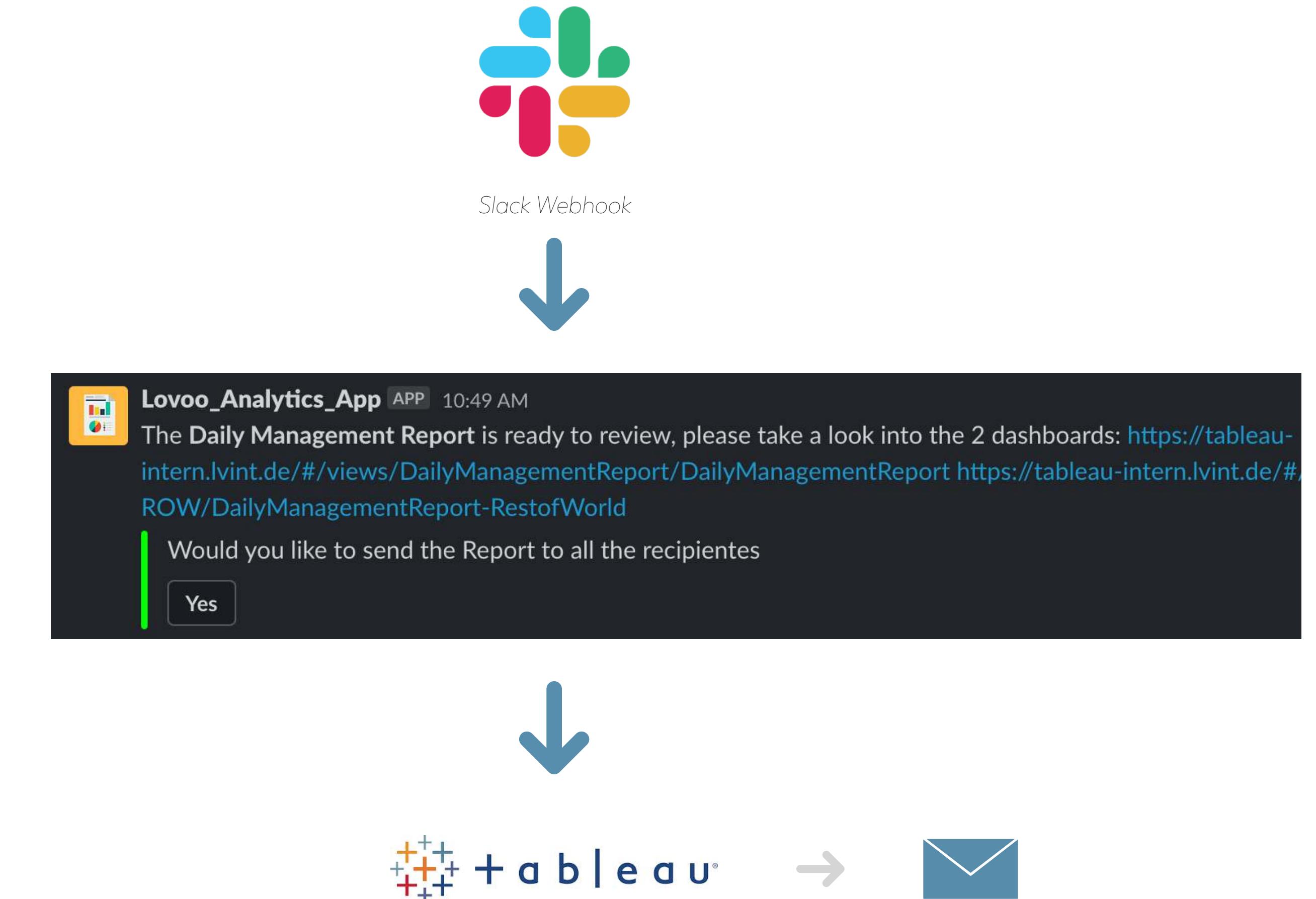
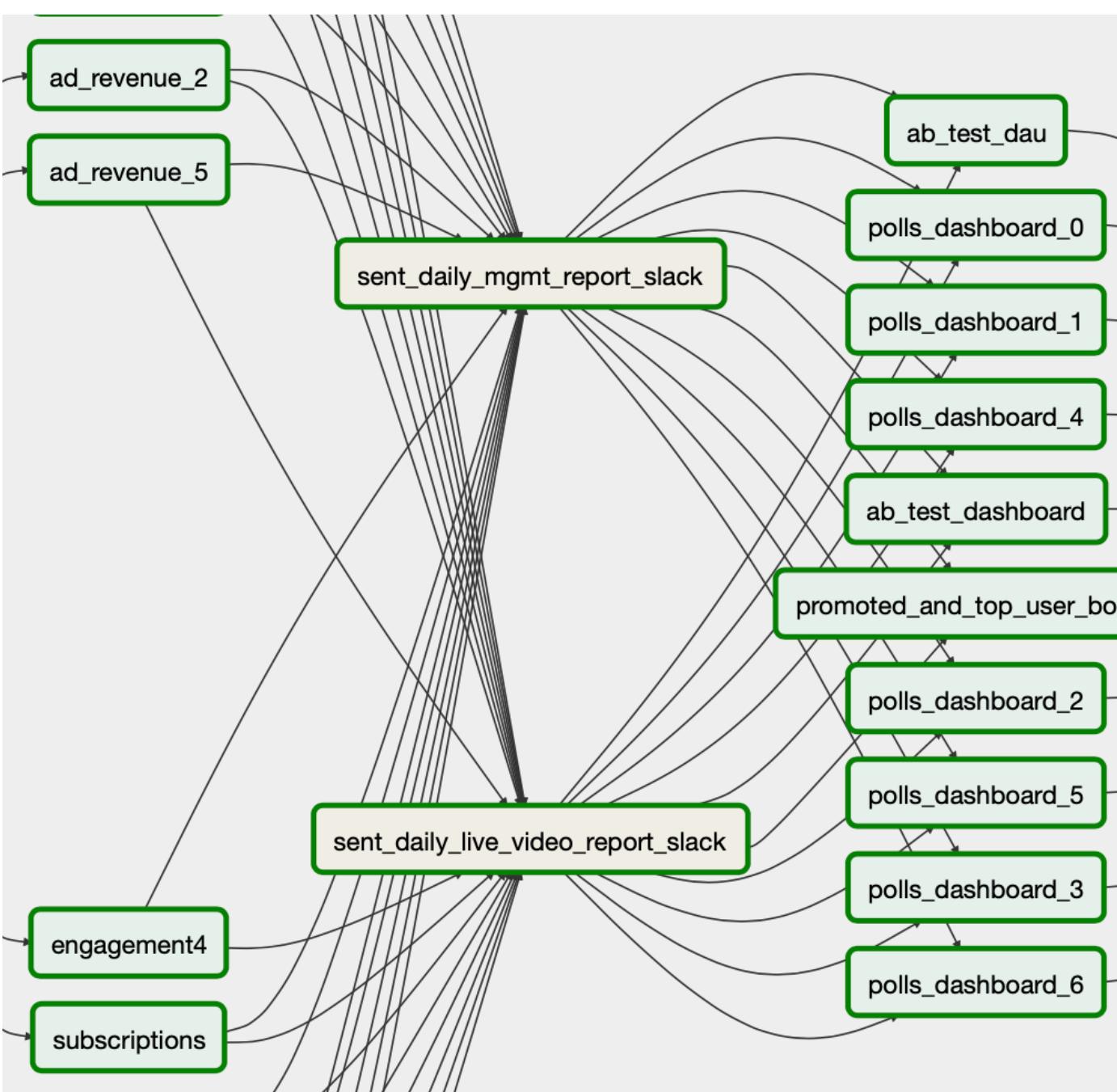
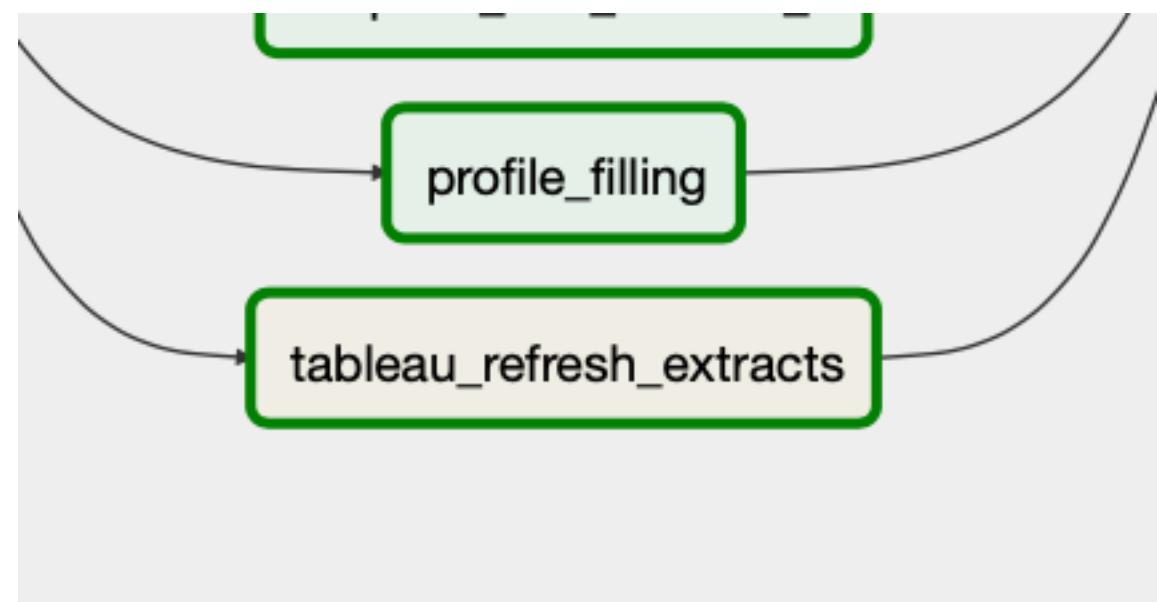


Tableau Extracts



```
bash_command
1 curl -X POST http://35.205.226.12:8007
```

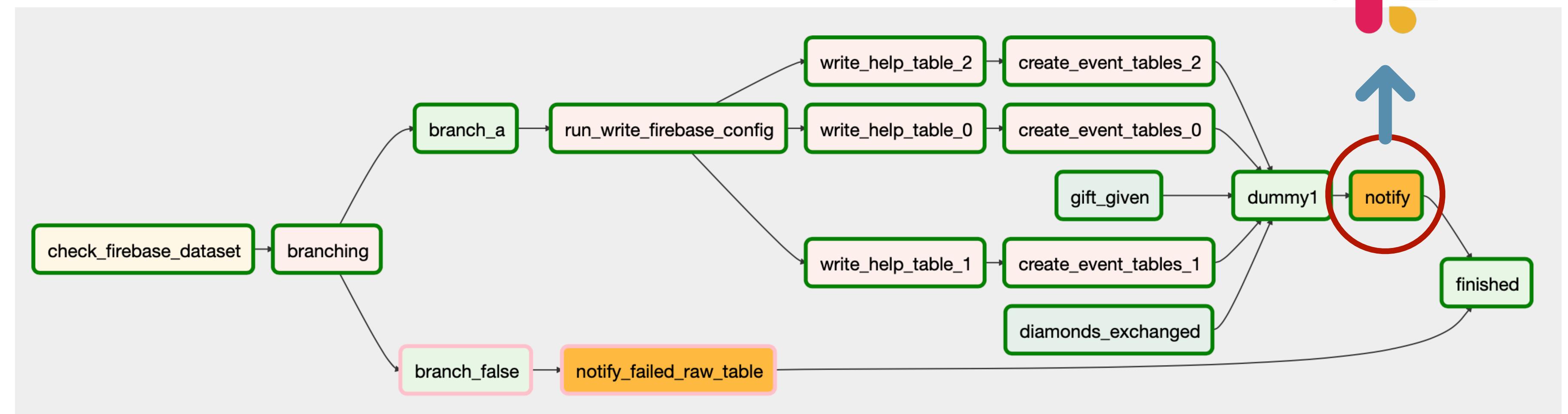


 + a b l e a u°

```
6 $tabcmd runschedule "Daily Extract Refreshes (9:00 AM)"
```

Is Airflow finished?

by the way, this is branching...



Is Airflow finished?



```

airflow-bot APP 6:47 AM
bi_firestore_live_events: Finished

airflow-bot APP 6:53 AM
bi_firestore_importer: Finished

airflow-bot APP 7:04 AM
1/2 analytics_jobs_live-1_user_level_live: Finished
2/2 analytics_jobs_live-2_agg_level_live: Finished
Analytics Live Pipeline Completed analytics_jobs_live: Finished

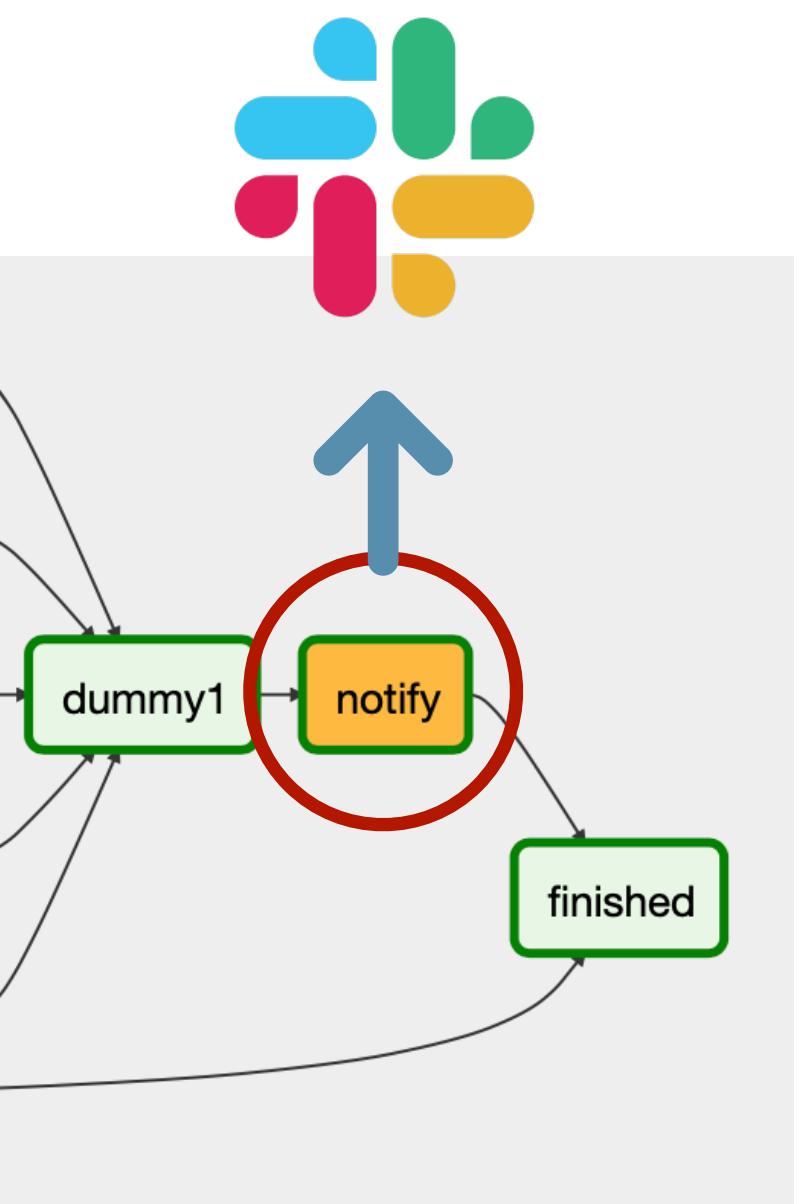
airflow-bot APP 7:21 AM
1/2 analytics_jobs-1_user_level: Finished

airflow-bot APP 8:55 AM
bi_marketing_events_jobs: Finished

airflow-bot APP 9:21 AM
2/2 analytics_jobs-2_agg_level: Finished
Analytics Pipeline Completed analytics_jobs: Finished

airflow-bot APP 1:26 PM
AppSumer Lovoo: Finished @piotr.predkiewicz

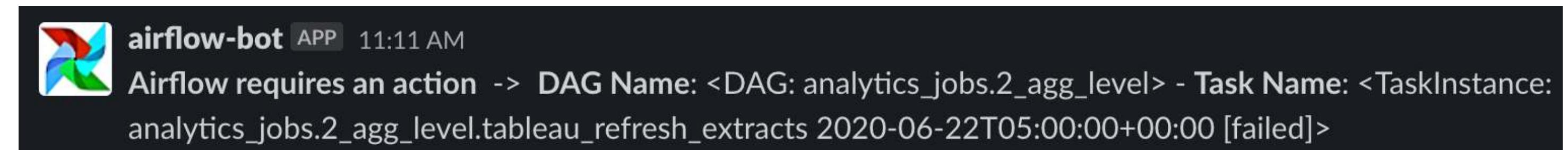
airflow-bot APP 2:35 PM
AppSumer Hayi: Finished @piotr.predkiewicz
  
```



BECAUSE SH!]
HAPPENS!

Error Alerting

```
'on_failure_callback': on_failure_callback,
```



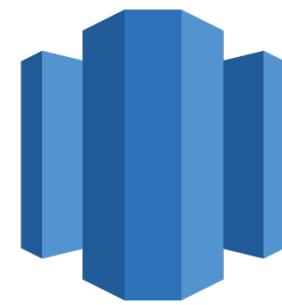
```
def on_failure_callback(context):
    operator = SlackAPIPostOperator(
        task_id='notify_fail',
        channel="#the_channel",
        token='your_Slack_bot_token',
        username='airflow-bot',
        text= str('*Airflow requires an action* {} Task: {}').format(
            str(context['dag']), str(context['task_instance'])))
    )
    return operator.execute(context=context)
```



Integrating Data Sources

this code belongs to the DAG.py file

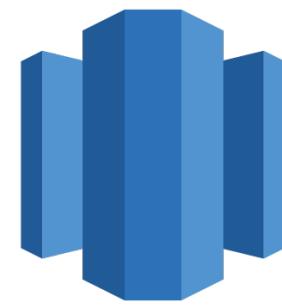
```
t1a = PythonOperator(  
    task_id='load_table_lovoo_transaction_groups_{}'.format(i),  
    python_callable=import_day_callable,  
    provide_context=True,  
    templates_dict={'exec_date': exec_date, 'table_name':'lovoo_transaction_groups'},  
    dag=dag)
```



Integrating Data Sources

this code belongs to the DAG.py file

```
from BI.redshift_importer import import_datalake_redshift_data
def import_day_callable(**kwargs):
    exec_date = kwargs.get('templates_dict').get('exec_date')
    table_name = kwargs.get('templates_dict').get('table_name')
    return import_datalake_redshift_data(table_name,
                                         'load_job_dataframe_to_bq',
                                         exec_date=exec_date)
```



Integrating Data Sources

this code belongs to the importer.py file

```
def postgreSQL_connection():
    try:
        # Using a Hook for getting the Redshift credentials from the Airflow connections -
        connection = BaseHook.get_connection("redshift_tmg")
        password = connection.password
        host = connection.host
        dbname = connection.schema
        user = connection.login
        port = connection.port

        conn = psycopg2.connect("dbname='{}' user='{}' port='{}' host='{}' password='{}'".format(dbname, user, port, host, password))
        cursor = conn.cursor()

    except Exception as e:
        print("I am unable to connect to the database: " + str(e))

    return cursor,conn
```



Integrating Data Sources

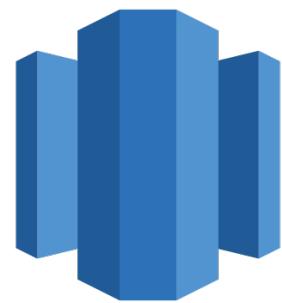
this pseudo-code belongs to the importer.py file

```
def import_datalake_redshift_data(table_name, method_type, exec_date, **kwargs):
    # Cursor & Connection
    cursor, conn = postgreSQL_connection()

    - Create dynamically a SQL query using the input parameters table_name and exec_date
    query = "select * from a_datalake.{} where data_updated_at::date >= '{}'".format(table_name, exec_date)

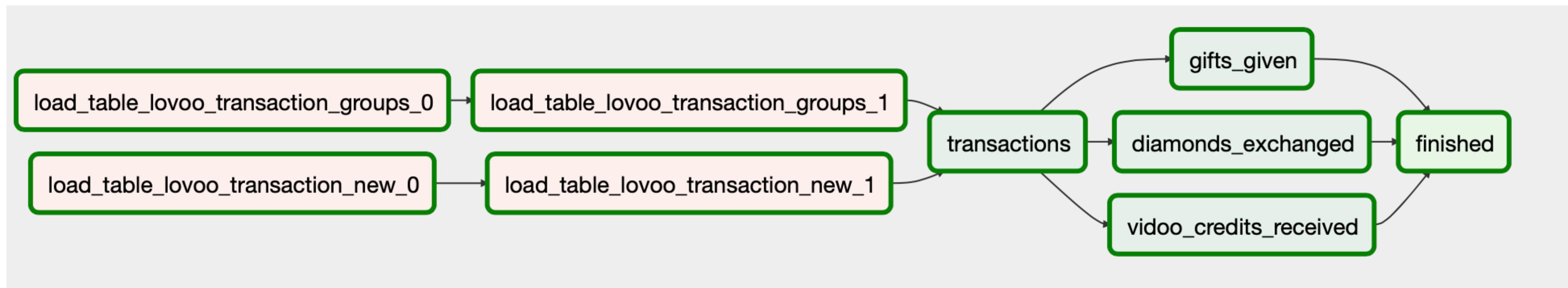
    - Use the query to request the data using the cursor
    cursor.execute(query)

    - use any method to upload the data to BigQuery
    df = cursor.fetchall()
    df = pd.DataFrame(df)
    job = client.load_table_from_dataframe(
        df, table_name, job_config=job_config
    )
    return whether it was successful or not
```



Integrating Data Sources

2 Tables - 2 Days -> ELT in BQ



Data Importers

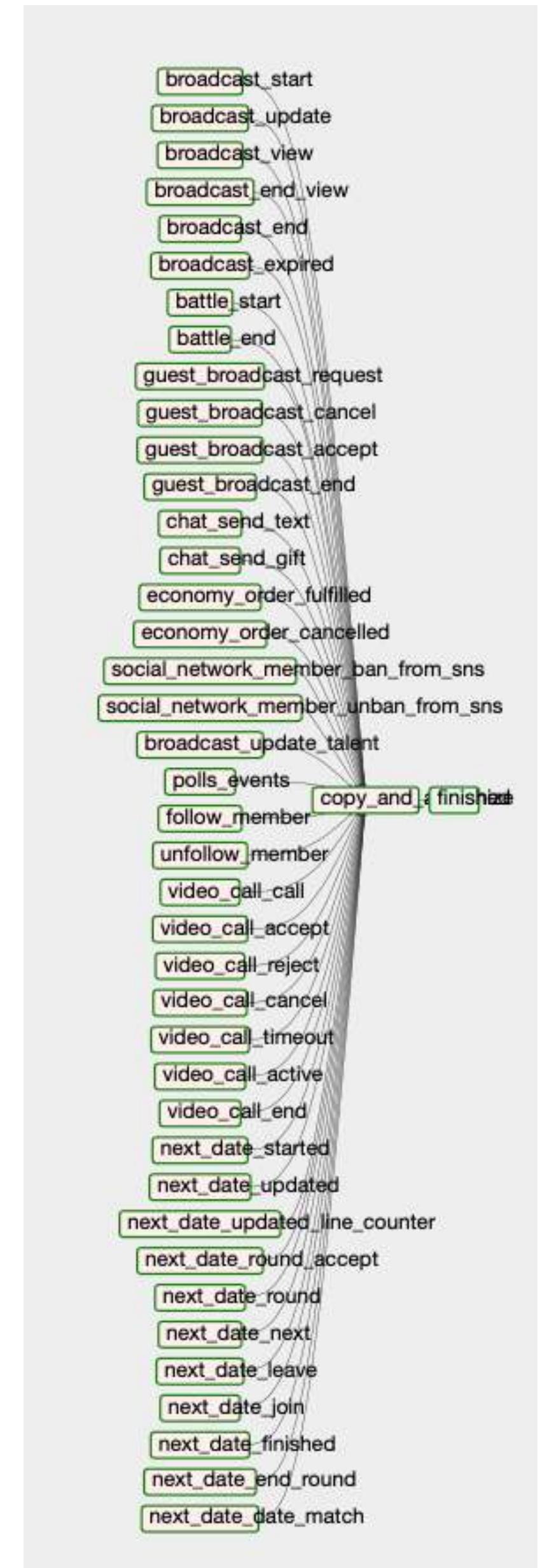
- Redshift
- Firebase (very dynamic)
- Google Cloud Storage (Adjust, Merger)
- Appsumer, Shopify, Paypal, AppStore, Adyen
- S3 Storage

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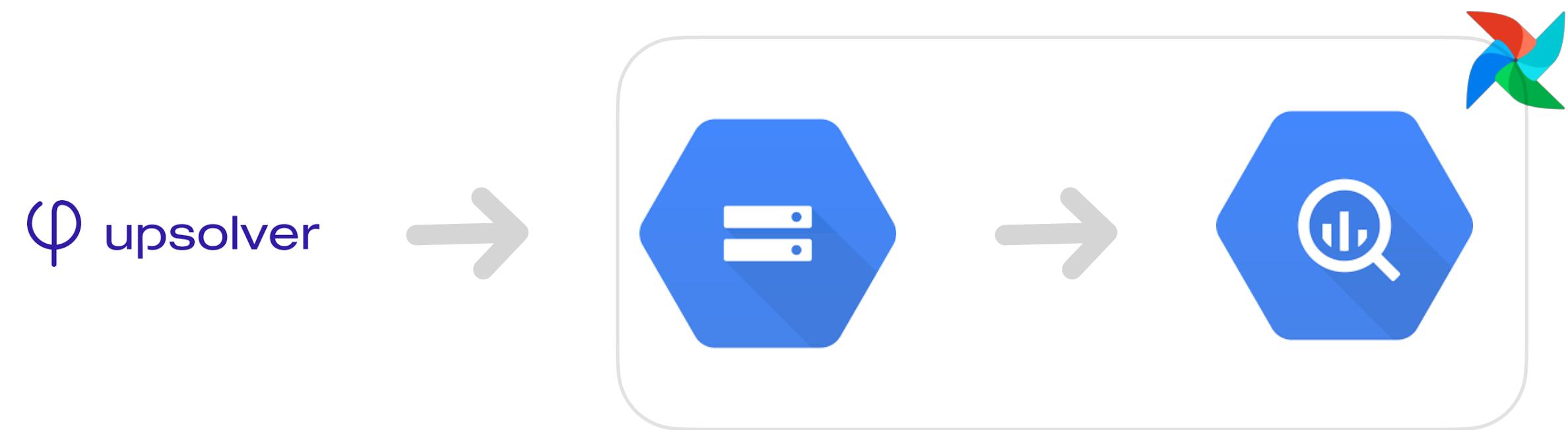
YES, VERY DYNAMIC...

Creating Tasks Dynamically



Creating Tasks Dynamically

1. Creating a plain text with meaningful structure
2. Create a task based on a PythonOperator
3. Define and write your Callable (your custom code)



Creating Tasks Dynamically

JSON File

```
{  
    "broadcast_start": {  
        "id": "81ce53e3-2ca6-48a2-88f7-493f7fc2c364",  
        "format": "json"  
    },  
    "broadcast_update": {  
        "id": "de7cbe04-9b53-4b4c-bc18-4d065ed3830e",  
        "format": "json"  
    },  
    "broadcast_view": {  
        "id": "2a4a0093-baee-47a2-8817-aebdb469b1b1",  
        "format": "json"  
    },  
    "broadcast_end_view": {  
        "id": "3af65f80-94a3-42c4-8ed9-502134605d27",  
        "format": "json"  
    },  
    "broadcast_end": {  
        "id": "37fa9fdf-8384-41a7-83aa-1263814b3585",  
        "format": "json"  
    },  
    ...  
}
```

Creating Tasks Dynamically

this code belongs to the DAG.py file

```
# Iterates over all the Mapping file and extracts the event name for generating all the task-events
for event_name in event_mapping:
    event_name_task = PythonOperator(
        task_id=str(event_name),
        provide_context=True,
        python_callable=run_import_day,
        templates_dict={'exec_date': exec_date, 'event_name': event_name,
                        'dataset': dataset, 'bucket_name': bucket_name},
        dag=dag)
```

Creating Tasks Dynamically

this code belongs to the DAG.py file

```
# Function that will be called by the Python operator and will write a table partition in BQ
def run_import_day(**kwargs):
    dataset = 'events_input_analytics_tmgtbackend'
    bucket_name = 'lovoo-tmgt-transfer'
    import_gcs_to_bq(exec_date=kwargs.get('templates_dict').get('exec_date'),
                      event_name=kwargs.get('templates_dict').get('event_name'),
                      dataset=kwargs.get('templates_dict').get('dataset'),
                      bucket_name=kwargs.get('templates_dict').get('bucket_name'),
    )
```

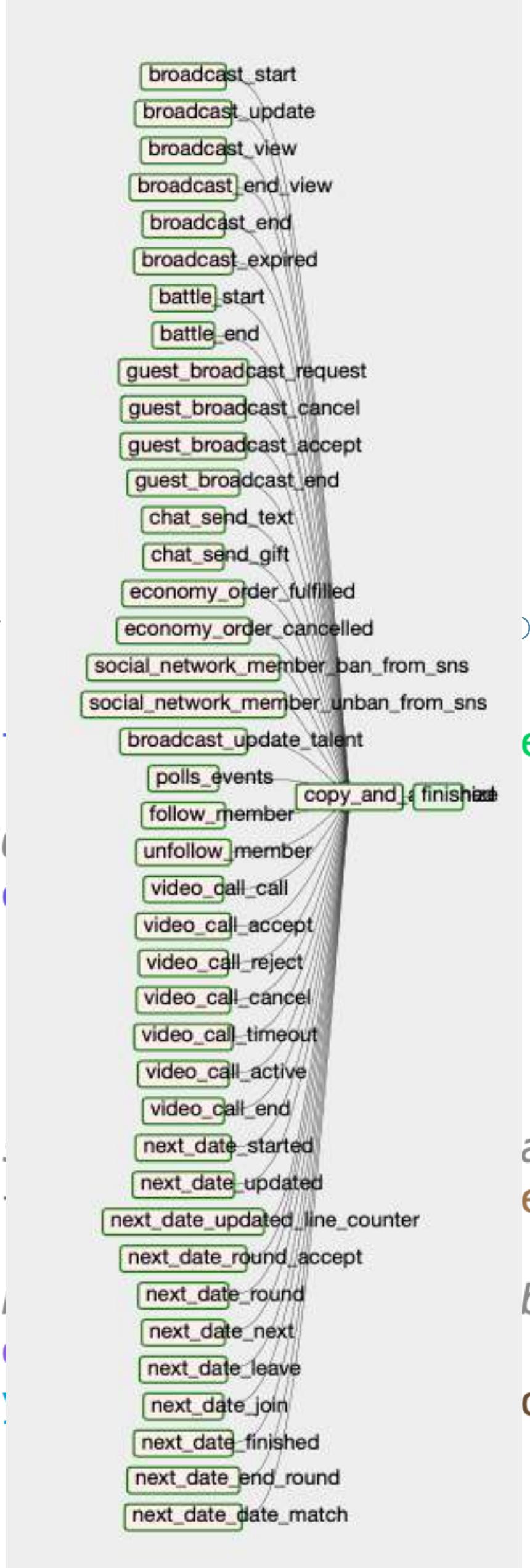
Creating Tasks Dynamically

this is your custom code (Pseudo-Code)

```
def import_gcs_to_bq(exec_date, event_name, dataset, bucket_name,**op_kwargs):  
  
    # read the structured JSON file  
    event_mapping = json.load(read_file)  
  
    # mapping the id and the event_name  
    id_event = event_mapping[event_name]['id']  
  
    # gathering the blobs inside the bucket - array of paths  
    path_array.append('gs://{}//{}//exec_date_file.json'.format(bucket_name, id_event))  
  
    # BigQuery Job to Load the JSON files to a table  
    load_job = bq_client.load_table_from_uri(  
        tuple(path_array), table_dest, job_config=job_config  
    )
```

Creating Tasks Dynamically

```
def import_gcs_to_bq(exec_date, bucket_name, **op_kwargs):  
  
    # read the structured JSON file  
    event_mapping = json.load(open('event_mapping.json'))  
  
    # mapping the id and the event  
    id_event = event_mapping[exec_date]  
  
    # gathering the blobs in the bucket  
    path_array.append('gs://' + bucket_name + '/' + id_event)  
  
    # BigQuery Job to Load the data  
    load_job = bq_client.load_table_from_uri(  
        tuple(path_array),  
        job_config=job_config)
```



this is the pseudo-Code)
et, bucket_name, **op_kwargs):

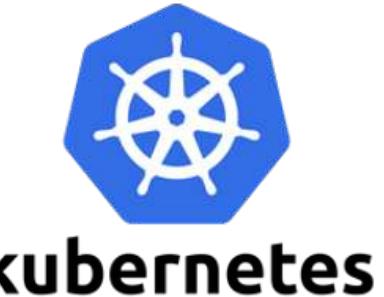
ay of paths
e.json'.format(bucket_name, id_event))
ble
onfig=job_config

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Recap and Conclusion

```
return KubernetesPodOperator(  
    startup_timeout_seconds=60 * 10, # we need seconds here as int, 10min now  
    task_id= 'appsumer_import_' + iso_date.replace('-', '_'),  
    namespace='default',  
    image=task_kwargs.get('image'),  
    cmd=task_kwargs.get('command'),  
    secrets=[appsumer_pass, service_account],  
    env_vars=env_vars,  
    name=task_kwargs.get('name'),  
    is_delete_operator_pod=True,  
    dag=dag,  
    dt=dt,  
    pool="appsumer_pool",  
    get_logs=True,  
    resources=resources,  
    affinity={  
        'nodeAffinity': {  
            # requiredDuringSchedulingIgnoredDuringExecution means in order  
            # for a pod to be scheduled on a node, the node must have the  
            # specified labels. However, if labels on a node change at  
            # runtime such that the affinity rules on a pod are no longer  
            # met, the pod will still continue to run on the node.  
        'requiredDuringSchedulingIgnoredDuringExecution': {  
            'nodeSelectorTerms': [{  
                'matchExpressions': [{  
                    'key': 'kuberunoperator',  
                    'operator': 'In',  
                    'values': [  
                        'true',  
                    ]  
                }]
```



Recap and Conclusion

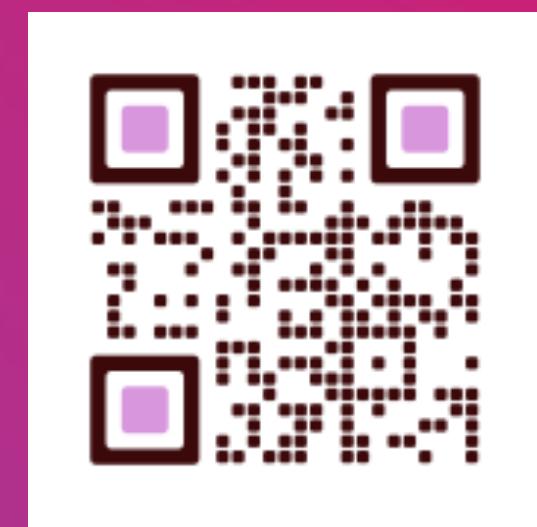
- Using an Alpha version (Google Composer) in Production was challenging!
- Focus on what's important - Google Cloud Composer
- Airflow leverages a bunch of Operators OOTB
- Always room for improvement
- No magic recipe to use - stay flexible

Gracias.

July 16, 2020 Berlin - Germany

Feedback and Questions

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