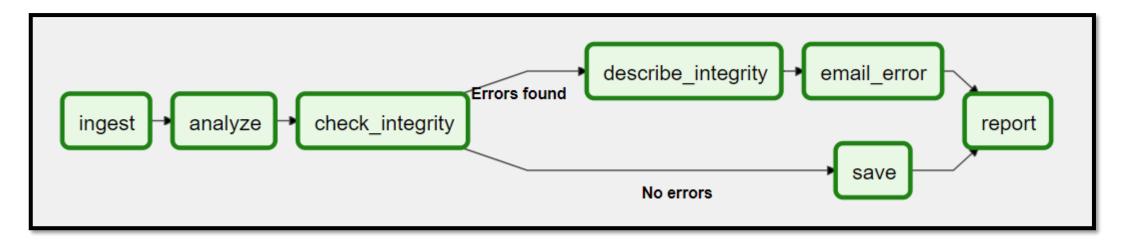
What is Airflow?





Apache Airflow

- · Airflow is a platform, **built with Python**, that lets you build and run workflows, **programmatically using Python**.
 - · A workflow is represented as a DAG (a Directed Acyclic Graph)
 - · DAG contains individual tasks, arranged with dependencies
- · Directed Acyclic Graph is a specialized version of a Directed Graph which contains no cycles (i.e., no loops)



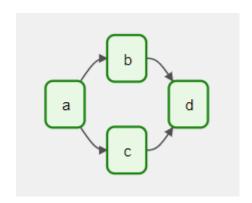


DAG

- · A *DAG* (Directed Acyclic Graph) is the core concept of Airflow, collecting <u>Tasks</u> together, organized with dependencies and relationships to say how they should run.
- This simple DAG defines four Tasks A, B, C, and D and dictates:
 - the order in which they have to run, and
 - which tasks depend on what other tasks.
- It will also say how often to run the DAG maybe "every 5 minutes starting tomorrow", or "every day since January 1st, 2020".

Note:

 The DAG itself doesn't care about what is happening inside the tasks; it is merely concerned with how to execute them - the order to run them in, how many times to retry them, if they have timeouts, and so on



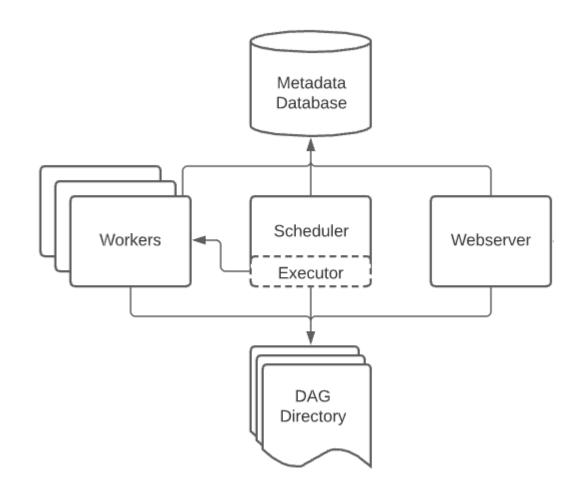
Core Components





Airflow Components

- A <u>scheduler</u>, which handles both triggering scheduled workflows, and submitting <u>Tasks</u> to the executor to run.
- An <u>executor</u>, which handles running tasks.
- A webserver, for Airflow UI to inspect, trigger and debug the behaviour of DAGs and tasks.
- A folder of DAG files, read by the scheduler and executor (and any workers the executor has)
- A metadata database, used by the scheduler, executor and webserver to store state.



Airflow UI







DAGs

Security

Browse

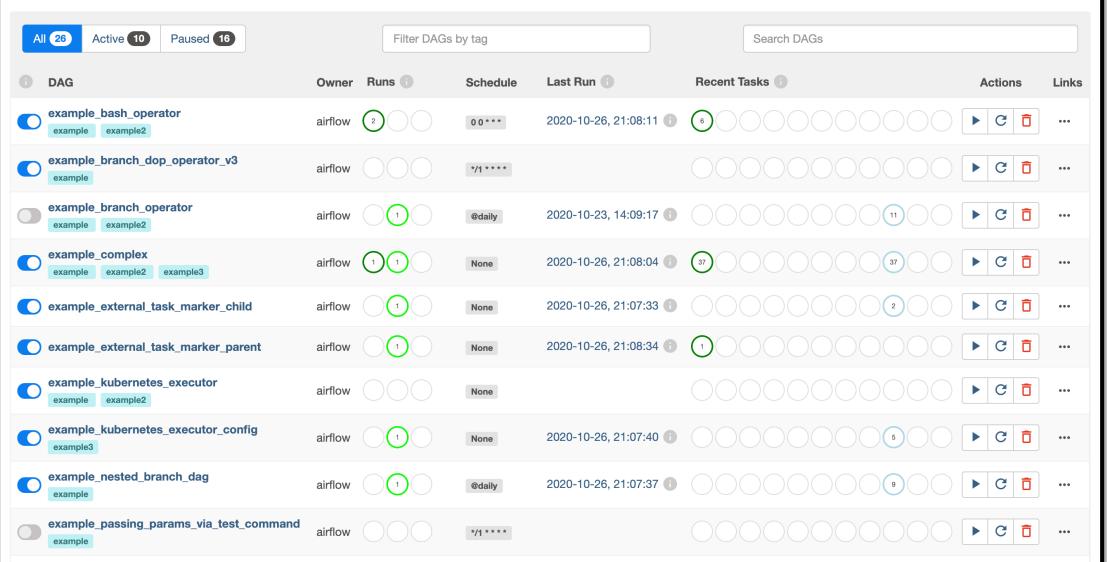
Admin

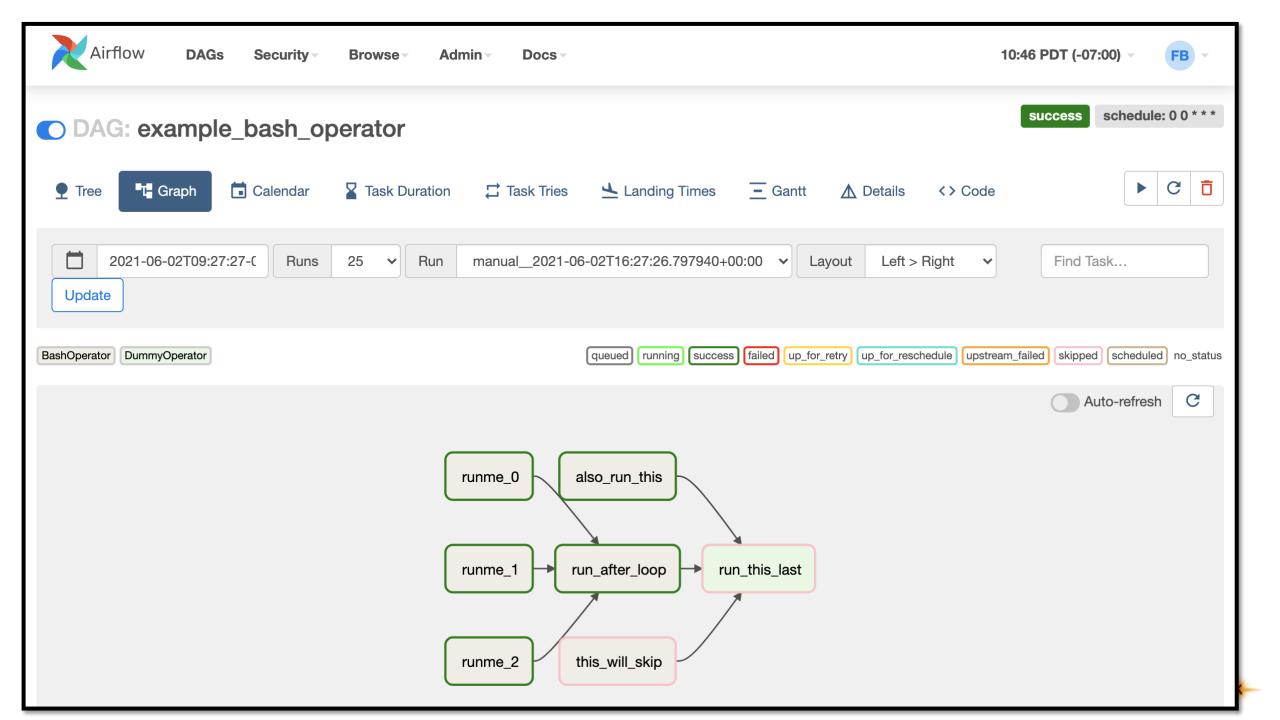
Docs

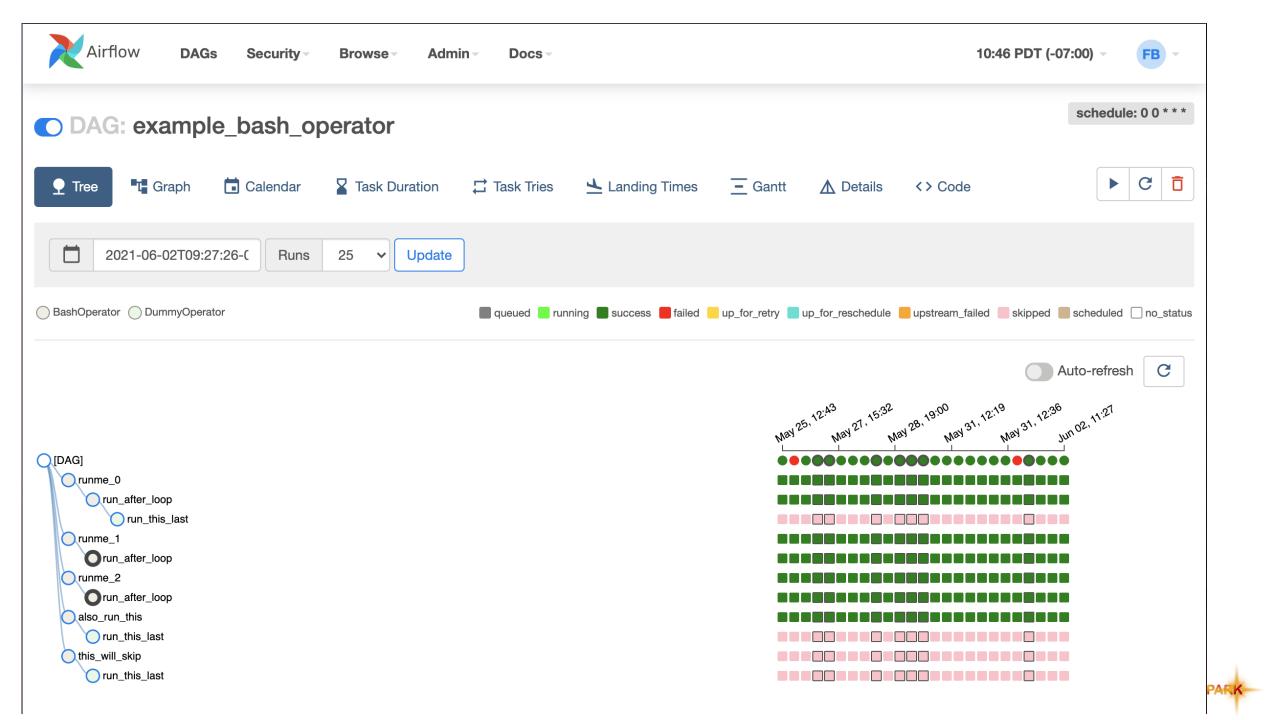


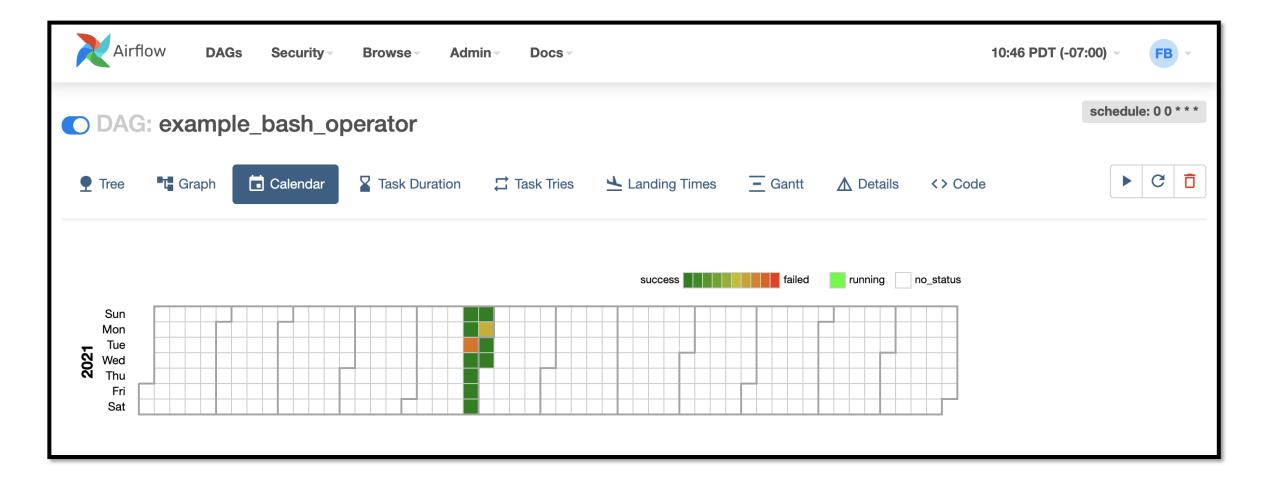


DAGs









Building DAGs





```
dag = DAG(
    'tutorial',
    default_args=default_args,
    description='A simple tutorial DAG',
    schedule_interval=timedelta(days=1),
    start_date=datetime(2021, 1, 1),
)
```

```
default args = {
    'owner': 'airflow',
    'depends on past': False,
    'email': ['airflow@example.com'],
    'email on failure': False,
    'email on retry': False,
    'retries': 1,
    'retry delay': timedelta(minutes=5),
    # 'queue': 'bash queue',
    # 'pool': 'backfill',
    # 'priority weight': 10,
    # 'end_date': datetime(2016, 1, 1),
    # 'wait for downstream': False,
    # 'dag': dag,
    # 'sla': timedelta(hours=2),
    # 'execution timeout': timedelta(seconds=300),
    # 'on failure callback': some function,
    # 'on_success_callback': some_other_function,
    # 'on retry callback': another function,
    # 'sla miss_callback': yet_another_function,
    # 'trigger rule': 'all success'
```

DAG Runs (schedule_intervals)

- Each DAG may or may not have a schedule (**why not?**), which informs how DAG Runs are created.
 - **schedule_interval** is defined as a DAG argument, and receives preferably a cron expression as a **str**, or a **datetime.timedelta** object.

preset	meaning	cron
None	Don't schedule, use for exclusively "externally triggered" DAGs	
@once	Schedule once and only once	
@hourly	Run once an hour at the beginning of the hour	0 * * *
@daily	Run once a day at midnight	00***
@weekly	Run once a week at midnight on Sunday morning	00**0
@monthly	Run once a month at midnight of the first day of the month	001**
@quarterly	Run once a quarter at midnight on the first day	0 0 1 */3 *
@yearly	Run once a year at midnight of January 1	0011*

Crontabguru (https://crontab.guru/)



*	any value
,	value list separator
-	range of values
/	step values
@yearly	(non-standard)
@annually	(non-standard)
@monthly	(non-standard)
@weekly	(non-standard)
@daily	(non-standard)
@hourly	(non-standard)
@reboot	(non-standard)

Operators

- · While DAGs describe how to run a workflow, **Operators** determine what actually gets done by a task.
- The DAG will make sure that operators run in the correct order
 - · Once the dependencies are met, operators generally run independently.
 - In fact, they may run on two completely different machines.
 - · If two operators need to share information, (e.g., small amount data, variable name, etc), consider combining them into a single operator.
 - · If it absolutely can't be avoided, Airflow does have a feature for operator cross-communication called XCom

Operators (cont...)

Airflow provides operators for many common tasks, including:

- BashOperator executes a bash command
- PythonOperator calls an arbitrary Python function
- EmailOperator sends an email
- SimpleHttpOperator sends an HTTP request
- MySqlOperator, SqliteOperator, PostgresOperator, MsSqlOperator,
 OracleOperator, JdbcOperator, etc. executes a SQL command
- Sensor an Operator that waits (polls) for a certain time, file, database row, S3 key, etc...

In addition to these basic building blocks, there are many more specific operators:

DockerOperator, HiveOperator, S3FileTransformOperator, PrestoToMySqlTransfer, SlackAPIOperator ... you get the idea!

Airflow Tasks

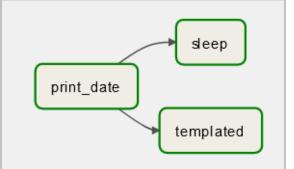
- · A Task is the basic unit of execution in Airflow
 - · Tasks instantiate Airflow Operators or Sensors (Triggers) to execute predefined functionality
 - Tasks are arranged (i.e., added/inserted) into <u>DAGs</u>

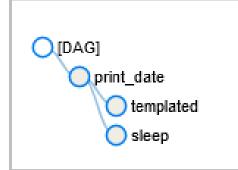
```
task_print_date = BashOperator(
   task_id='print_date',
   bash_command='date',
   dag = dag,
)
```

```
templated_task = BashOperator(
    task_id='templated',
    depends_on_past=False,
    bash_command=templated_command,
    params={'my_param': 'Parameter I passed in'},
    dag = dag,
)
```

- · Upstream and downstream dependencies are set between the tasks into order to express the order they should run in.
 - · First declare the tasks, then declare its dependencies

```
task_print_date >> [task_sleep, templated_task]
```







Sensors

- · Sensors are a special type of Operator that are designed to do exactly one thing wait for something to occur.
- · It can be **time-based**, or **waiting for a file**, or an **external event**, but all they do is wait until something happens, and then succeed so their downstream tasks can run.

https://airflow.apache.org/docs/apache-airflow/stable/_api/airflow/sensors/index.html

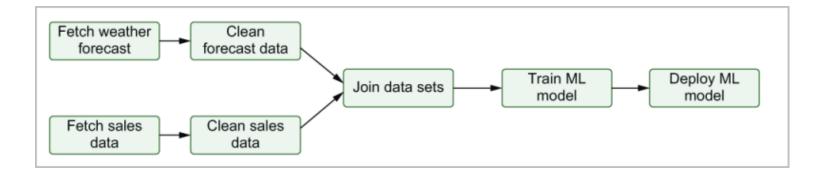
- airflow.sensors.base
- airflow.sensors.base_sensor_operator
- airflow.sensors.bash
- airflow.sensors.date_time
- airflow.sensors.date_time_sensor
- airflow.sensors.external_task
- airflow.sensors.external_task_sensor
- airflow.sensors.filesystem
- airflow.sensors.hdfs_sensor
- airflow.sensors.hive_partition_sensor
- airflow.sensors.http_sensor
- airflow.sensors.metastore_partition_sensor
- airflow.sensors.named_hive_partition_sensor

- airflow.sensors.python
- airflow.sensors.s3_key_sensor
- airflow.sensors.s3_prefix_sensor
- airflow.sensors.smart_sensor
- airflow.sensors.sql
- airflow.sensors.sql_sensor
- airflow.sensors.time_delta
- airflow.sensors.time_delta_sensor
- airflow.sensors.time_sensor
- airflow.sensors.web_hdfs_sensor
- airflow.sensors.weekday

DAG Workflows







· Linear Dependency: 2 Workflow Path executing in parallel:

```
fetch_weather_forecast >> clean_weather_forecast
fetch_sales_data >> clean_sales_data
```

Fan-out: One-2-Many Dependency (implicit in our case)
start = DummyOperator(task_id="start")

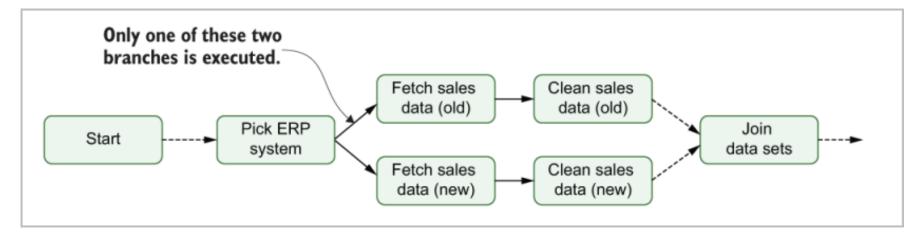
```
start >> [fetch_weather_forecast, fetch_sales_data]
```

· Fan-in: Multiple-2-One Dependency (implicit in our case)

```
[clean_weather_forecast, clean_sales_data] >> join_data_sets
```

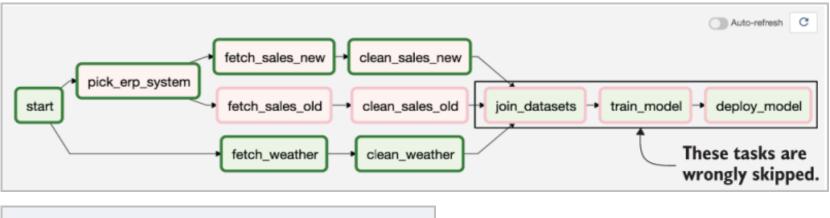


Branching

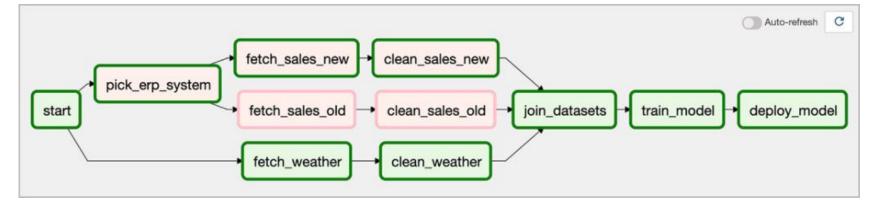


```
def _pick_erp_system(**context):
    if context["execution_date"] < ERP_SWITCH_DATE:</pre>
        return "fetch_sales_old"
    else:
        return "fetch_sales_new"
pick_erp_system = BranchPythonOperator(
    task_id="pick_erp_system",
    python callable= pick erp system,
pick erp system >> [fetch sales old, fetch sales new]
```

Trigger Rules



trigger_rule = "all_success"



Trigger Rules (cont...)

- all_success (default): All upstream tasks have succeeded
- all_failed: All upstream tasks are in a failed or upstream_failed state
- all_done : All upstream tasks are done with their execution
- one_failed: At least one upstream task has failed (does not wait for all upstream tasks to be done)
- one_success: At least one upstream task has succeeded (does not wait for all upstream tasks to be done)
- none_failed: All upstream tasks have not failed or upstream_failed that is, all upstream tasks have succeeded or been skipped
- none_failed_min_one_success: All upstream tasks have not failed or upstream_failed, and at least one upstream task has succeeded.
- none_skipped: No upstream task is in a skipped state that is, all upstream tasks are in a success, failed, or upstream_failed state
- always: No dependencies at all, run this task at any time

Task Cross-Communication (xcom)

- · Airflow xcom is used to "pass" data between tasks
 - Tasks are completely isolated otherwise
 - Guidance is that it should be used for small amounts of data.
 - Source Code implies that max xcom size is 48kb but is never used
 - https://airflow.apache.org/docs/apache-airflow/stable/_modules/airflow/models/xcom.html
 - · Seems to be dependent on backend database blob size:
 - Sqlite: default max of 1 GB (absolute maximum of 2 GB)
 - · Mysql: 64 kb
- xcom_push: pushes data into xcom
- · xcom_pull: pulls data from xcom

SLA

- · Service Level Agreements (SLAs), or the **time** by which a <u>task</u> or <u>DAG</u> should have succeeded, can be set at a task level as a timedelta.
- · If one or many instances have not succeeded by that time, an alert email is sent with the list of tasks that missed their SLA.
 - The SLA miss event is also recorded in the database and made available in the web UI under Browse->SLA Misses where events can be analyzed and documented.
- · SLAs can be configured for scheduled tasks by using the **sla** parameter.
 - · In addition to sending alerts, the **sla_miss_callback** specifies an additional method to be invoked when the SLA is not met to invoke any custom functionality



Providers Packages

• Airbyte	• Elasticsearch	• Oracle
• Alibaba	• Exasol	Pagerduty
• Amazon	• Facebook	 Papermill
Apache Beam	• File Transfer Protocol (FTP)	• Plexus
Apache Cassandra	• Google	 PostgreSQL
• Apache Drill	• gRPC	Presto
• Apache Druid	• Hashicorp	Qubole
Apache HDFS	 Hypertext Transfer Protocol (HTTP) 	• Redis
• Apache Hive	• Influx DB	 Salesforce
Apache Kylin	• Internet Message Access Protocol (IMAP)	• Samba
Apache Livy	• Java Database Connectivity (JDBC)	 Segment
Apache Pig	• Jenkins	 Sendgrid
• Apache Pinot	• Jira	• SFTP
Apache Spark	Microsoft Azure	Singularity
Apache Sqoop	• Microsoft PowerShell Remoting Protocol	Slack
Asana	(PSRP)	Snowflake
• Celery	 Microsoft SQL Server (MSSQL) 	• SQLite
• IBM Cloudant	 Windows Remote Management (WinRM) 	• SSH
• Kubernetes	MongoDB	 Tableau
• Databricks	MySQL	• Telegram
• Datadog	• Neo4J	• Trino
• Dingding	• ODBC	 Vertica
• Discord	OpenFaaS	Yandex
• Docker	• Opsgenie	 Zendesk