
Kedro Airflow K8S Plugin

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GetInData

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INTRODUCTION

1.1 What is Airflow?

[Airflow](#) is a platform to programmatically author, schedule and monitor workflows. Workflows are represented as DAGs. Each DAG is represented by nodes, that define job to be executed. The DAGs are stored in the file storage, allowing user to run the pipeline once or schedule the recurring run.

1.2 What is Kubernetes?

[Kubernetes](#) is a platform for managing containerized workloads and services, that facilitates both declarative configuration and automation.

1.3 Why to integrate Kedro project with Airflow nad Kubernetes?

Airflow's main attitude is the portability. Once you define a pipeline, it can be started on any Kubernetes cluster. The code to execute is stored inside docker images that cover not only the source itself, but all the libraries and entire execution environment. Portability is also one of key Kedro aspects, as the pipelines must be versionable and packageable. Kedro, with [Kedro-docker](#) plugin do a fantastic job to achieve this and Airflow looks like a nice add-on to run the pipelines on powerful remote Kubernetes clusters.

INSTALLATION

2.1 Installation guide

2.1.1 Kedro setup

First, you need to install base Kedro package in <17.0 version

Kedro 17.0 is supported by kedro-airflow-k8s, but not by kedro-mlflow yet, so the latest version from 0.16 family is recommended.

```
$ pip install 'kedro<0.17'
```

2.1.2 Plugin installation

Install from PyPI

You can install kedro-airflow-k8s plugin from PyPi with pip:

```
pip install --upgrade kedro-airflow-k8s
```

Install from sources

You may want to install the develop branch which has unreleased features:

```
pip install git+https://github.com/getindata/kedro-airflow-k8s.git@develop
```

2.1.3 Available commands

You can check available commands by going into project directory and running:

```
$ kedro airflow-k8s

Usage: kedro airflow-k8s [OPTIONS] COMMAND [ARGS]...

Options:
  -e, --env TEXT  Environment to use.
  -h, --help      Show this message and exit.
```

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Commands:	
<code>compile</code>	Create an Airflow DAG for a project
<code>init</code>	Initializes configuration for the plugin
<code>list-pipelines</code>	List pipelines generated by this plugin
<code>run-once</code>	Uploads pipeline to Airflow and runs once
<code>schedule</code>	Uploads pipeline to Airflow with given schedule
<code>ui</code>	Open Apache Airflow UI in new browser tab
<code>upload-pipeline</code>	Uploads pipeline to Airflow DAG location

compile

`compile` command takes one argument, which is the directory name containing configuration (relative to `conf` folder). As an outcome, `dag` directory contains python file with generated DAG.

init

`init` command adds default plugin configuration to the project, based on Apache Airflow CLI input. It also allows optionally adding github actions, to streamline project build and upload.

list-pipelines

`list-pipelines` lists all pipelines generated by this plugin which exist in Airflow server. All generated DAGs are tagged with tag `generated_with_kedro_airflow_k8s:$PLUGIN_VERSION` and the prefix of this tag is used to distinguish among the other tags.

run-once

`run-once` command generates DAG from the pipeline, uploads it Airflow DAG location and triggers the DAG run as soon as the new DAG instance is available. It optionally allows waiting for DAG run completion, checking if success status is returned.

schedule

`schedule` command takes three arguments, one is the directory name containing configuration (relative to `conf` folder), the second one is the output location of generated dag, the third is cron like expression that relates to Airflow DAG `schedule_interval`.

ui

`ui` simplifies access to Apache Airflow console. It also allows open UI for the specific DAG.

upload-pipeline

upload-pipeline command takes two arguments, one is the directory name containing configuration (relative to conf folder), the second one is the output location of generated dag.

2.2 Configuration

Plugin maintains the configuration in the conf/base/airflow-k8s.yaml file.

```

# Base url of the Apache Airflow, should include the schema (http/https)
host: https://airflow.example.com

# Directory from where Apache Airflow is reading DAGs definitions
output: gs://airflow-bucket-example-com

# Configuration used to run the pipeline
run_config:

    # Name of the image to run as the pipeline steps
    image: airflow-k8s-plugin-demo

    # Pull policy to be used for the steps. Use Always if you push the images
    # on the same tag, or Never if you use only local images
    image_pull_policy: IfNotPresent

    # Pod startup timeout in seconds - if timeout passes the pipeline fails, default_
    ↪to 600
    startup_time: 600

    # Namespace for Airflow pods to be created
    namespace: airflow

    # Name of the Airflow experiment to be created
    experiment_name: Airflow K8S Plugin Demo

    # Name of the dag as it's presented in Airflow
    run_name: airflow-k8s-plugin-demo

    # Apache Airflow cron expression for scheduled runs
    cron_expression: "@daily"

    # Optional pipeline description
    description: "Very Important Pipeline"

    # Optional volume specification
    volume:
        # Storage class - use null (or no value) to use the default storage
        # class deployed on the Kubernetes cluster
        storageclass: # default
        # The size of the volume that is created. Applicable for some storage
        # classes
        size: 1Gi
        # Access mode of the volume used to exchange data. ReadWriteMany is
        # preferred, but it is not supported on some environements (like GKE)
        # Default value: ReadWriteOnce
        #access_modes: [ReadWriteMany]

```

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```

# Flag indicating if the data-volume-init step (copying raw data to the
# fresh volume) should be skipped
skip_init: False
# Allows to specify fsGroup executing pipelines within containers
# Default: root user group (to avoid issues with volumes in GKE)
owner: 0
# If set to True, shared persistent volume will not be created at all and all
↳ other parameters under
# `volume` are discarded
disabled: False

# Optional resources specification
resources:
# Default configuration used by all nodes that do not declare the
# resource configuration. It's optional. If node does not declare the resource
# configuration, __default__ is assigned by default, otherwise cluster
↳ defaults
# will be used.
__default__:
# Optional labels to be put into pod node selector
labels:
# Labels are user provided key value pairs
node_pool_label/k8s.io: example_value
requests:
# Optional amount of cpu resources requested from k8s
cpu: "1"
Optional amount of memory resource requested from k8s
memory: "1Gi"
limits:
# Optional amount of cpu resources limit on k8s
cpu: "1"
# Optional amount of memory resource limit on k8s
memory: "1Gi"
# Other arbitrary configurations to use, for example to indicate some
↳ exception resources
huge_machines:
labels:
big_node_pool: huge.10x
requests:
cpu: "16"
memory: "128Gi"
limits:
cpu: "32"
memory: "256Gi"

```

2.2.1 Indicate resources in pipeline nodes

Every node declared in kedro pipelines is executed inside pod. Pod definition declares resources to be used based on provided plugin configuration and presence of the tag `resources` in kedro node definition.

If no such tag is present, plugin will assign `__default__` from plugin `resources` configuration. If no `__default__` is given in plugin `resources` configuration or no `resources` configuration is given, pod definition will not be given any information on how to allocate resources to pod, thus default k8s cluster values will be used.

```

# train_model node is assigned resources from `huge_machines` configuration, if no_
↪such configuration exists,
# `__default__` is used, and if __default__ does not exist, k8s cluster default_
↪values are used
node(func=train_model, inputs=["X_train", "y_train"], outputs="regressor", name=
↪'train_model', tags=['resources:huge_machines'])
# evaluate_model node is assigned resources `__default__` configuration and if it_
↪does not exist,
# k8s cluster default values are used
node(func=evaluate_model, inputs=["X_train", "y_train"], outputs="regressor", name=
↪'evaluate_model')

```

2.2.2 Dynamic configuration support

kedro-airflow-k8s contains hook that enables TemplatedConfigLoader. It allows passing environment variables to configuration files. It reads all environment variables following KEDRO_CONFIG_ pattern, which you can later inject in configuration file using \${name} syntax.

There are two special variables KEDRO_CONFIG_COMMIT_ID, KEDRO_CONFIG_BRANCH_NAME with support specifying default when variable is not set, e.g. \${commit_id|dirty}

GETTING STARTED

3.1 Quickstart

3.1.1 Preerequisites

Although the plugin does not perform deployment, it's recommended to have access to Airflow DAG directory in order to test run the generated DAG.

3.1.2 Install the toy project with Kedro Airflow K8S support

It is a good practice to start by creating a new virtualenv before installing new packages. Therefore, use `virtualenv` command to create new env and activate it:

```
$ virtualenv venv-demo
created virtual environment CPython3.8.5.final.0-64 in 145ms
creator CPython3Posix(dest=/home/mario/kedro/venv-demo, clear=False, no_vcs_
↳ ignore=False, global=False)
  seeder FromAppData(download=False, pip=bundle, setuptools=bundle, wheel=bundle,
↳ via=copy, app_data_dir=/home/mario/.local/share/virtualenv)
    added seed packages: pip==20.3.1, setuptools==51.0.0, wheel==0.36.2
  activators BashActivator,CShellActivator,FishActivator,PowerShellActivator,
↳ PythonActivator,XonshActivator
$ source venv-demo/bin/activate
```

Then, `kedro` must be present to enable cloning the starter project, along with the latest version of `kedro-airflow-k8s` plugin and `kedro-docker`.

```
$ pip install 'kedro<0.17' kedro-airflow-k8s kedro-docker
```

With the dependencies in place, let's create a new project:

```
$ kedro new --starter=git+https://github.com/getindata/kedro-starter-spaceflights.git
↳ --checkout allow_nodes_with_commas
Project Name:
=====
Please enter a human readable name for your new project.
Spaces and punctuation are allowed.
[New Kedro Project]: Airflow K8S Plugin Demo

Repository Name:
=====
Please enter a directory name for your new project repository.
```

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```
Alphanumeric characters, hyphens and underscores are allowed.
Lowercase is recommended.
[airflow-k8s-plugin-demo]:

Python Package Name:
=====
Please enter a valid Python package name for your project package.
Alphanumeric characters and underscores are allowed.
Lowercase is recommended. Package name must start with a letter or underscore.
[airflow_k8s_plugin_demo]:

Change directory to the project generated in ${CWD}/airflow-k8s-plugin-demo

A best-practice setup includes initialising git and creating a virtual environment,
↳ before running
`kedro install` to install project-specific dependencies. Refer to the Kedro
documentation: https://kedro.readthedocs.io/
```

TODO: switch to the official `spaceflights` starter after <https://github.com/quantumblacklabs/kedro-starter-spaceflights/pull/10> is merged

Finally, go the demo project directory and ensure that `kedro-airflow-k8s` plugin is activated:

```
$ cd airflow-k8s-plugin-demo/
$ kedro install
(...)
Requirements installed!
$ kedro airflow-k8s --help
```console
$ kedro airflow-k8s

Usage: kedro airflow-k8s [OPTIONS] COMMAND [ARGS]...

Options:
-e, --env TEXT Environment to use.
-h, --help Show this message and exit.

Commands:
compile Create an Airflow DAG for a project
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ui Open Apache Airflow UI in new browser tab
upload-pipeline Uploads pipeline to Airflow DAG location
```

### 3.1.3 Build the docker image to be used on Airflow K8S runs

First, initialize the project with `kedro-docker` configuration by running:

```
$ kedro docker init
```

This command creates a several files, including `.dockerignore`. This file ensures that transient files are not included in the docker image and it requires small adjustment. Open it in your favourite text editor and extend the section `# except the following` by adding there:

```
!data/01_raw
```

This change enforces raw data existence in the image. Also, one of the limitations of running the Kedro pipeline on Airflow (and not on local environment) is inability to use `MemoryDataSets`, as the pipeline nodes do not share memory, so every artifact should be stored as file. The `spaceflights` demo configures four datasets as in-memory, so let's change the behaviour by adding these lines to `conf/base/catalog.yml`:

```
X_train:
 type: pickle.PickleDataSet
 filepath: data/05_model_input/X_train.pickle
 layer: model_input

y_train:
 type: pickle.PickleDataSet
 filepath: data/05_model_input/y_train.pickle
 layer: model_input

X_test:
 type: pickle.PickleDataSet
 filepath: data/05_model_input/X_test.pickle
 layer: model_input

y_test:
 type: pickle.PickleDataSet
 filepath: data/05_model_input/y_test.pickle
 layer: model_input
```

Finally, build the image:

```
kedro docker build
```

When execution finishes, your docker image is ready. If you don't use local cluster, you should push the image to the remote repository:

```
docker tag airflow_k8s_plugin_demo:latest remote.repo.url.com/airflow_k8s_plugin_
↪demo:latest
docker push remote.repo.url.com/airflow_k8s_plugin_demo:latest
```

### 3.1.4 Setup GIT repository

Plugin requires project to be under git repository. Perform [repository initialization](#) and commit project files

### 3.1.5 Compile DAG

Plugin requires configuration to be present. It's best to use:

```
kedro airflow-k8s init --with-github-actions --output ${AIRFLOW_DAG_FOLDER} https://
↪airflow.url
```

This command creates configuration file in `conf/pipelines/airflow-k8s.yaml` with some custom values and reference to Airflow passed in arguments. It also creates some default github actions.

When using this command, pay attention that the configuration expects `commit_id` and `google_project_id` to be present. Set them up by setting environment variable `KEDRO_CONFIG_COMMIT_ID` and `KEDRO_CONFIG_GOOGLE_PROJECT_ID`.

Also mlflow configuration has to be set up (if required by the project) as described in [mlflow section](#).

Having configuration ready, type:

```
kedro airflow-k8s -e pipelines compile
```

This command compiles pipeline and generates DAG in `dag/airflow_k8s_plugin_demo.py`. This file should be copied manually into Airflow DAG directory, that Airflow periodically scans. After it appears in airflow console, it is ready to be triggered.

As an alternative, one can use the following:

```
kedro airflow-k8s -e pipelines upload-pipeline -o ${AIRFLOW_DAG_HOME}
```

in order to get DAG copied directly to Airflow DAG folder. Google Cloud Storage locations are also supported with `gcs://` or `gs://` prefix in the parameter (this requires plugin to be installed with `pip install kedro-airflow-k8s[gcp]`).

## 3.2 GCP AI Platform support

Google Cloud's AI Platform offers couple services that simplify Machine Learning tasks.

### 3.2.1 Using kedro with AI Platform Notebooks

[AI Platform Notebooks](#) provides an easy way to manage and host JupyterLab based data science workbench environment. What we've found out is that the default images provided by a service cause some dependency conflicts. To avoid this issues make sure you use isolated virtual environment, e.g. [virtualenv](#). New virtual environment can be created by simply invoking `python -m virtualenv venv` command.



### 3.3 Mlflow support

If you use [MLflow](#) and [kedro-mlflow](#) for the Kedro pipeline runs monitoring, the plugin will automatically enable support for:

- starting the experiment when the pipeline starts,
- logging all the parameters, tags, metrics and artifacts under unified MLFlow run.

To make sure that the plugin discovery mechanism works, add `kedro-mlflow` as a dependencies to `src/requirements.in` and run:

```
$ pip-compile src/requirements.in > src/requirements.txt
$ kedro install
$ kedro mlflow init
```

Then, adjust the `kedro-mlflow` configuration and point to the `mlflow` server by editing `conf/local/mlflow.yml` and adjusting `mlflow_tracking_uri` key. Then, build the image:

```
$ kedro docker build
```

And re-push the image to the remote registry.

If `kedro-mlflow` is not installed as dependency and configuration is not in place (missing `kedro mlflow init`), the `MLflow` experiment will not be initialized and available for pipeline tasks in Apache Airflow DAG.



## INDICES AND TABLES

- `genindex`
- `modindex`
- `search`