

# **Big Data Workflow Scheduler**

## **Introducing Apache DolphinScheduler**

Apache Software Foundation Member  
Apache DolphinScheduler PMC

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## About Me



# William Kwok

Apache Software Foundation Member

Apache IPMC Member

PMC of Apache DolphinScheduler

Mentor of Apache SeaTunnel(incubating)

Founder of ClickHouse China Community

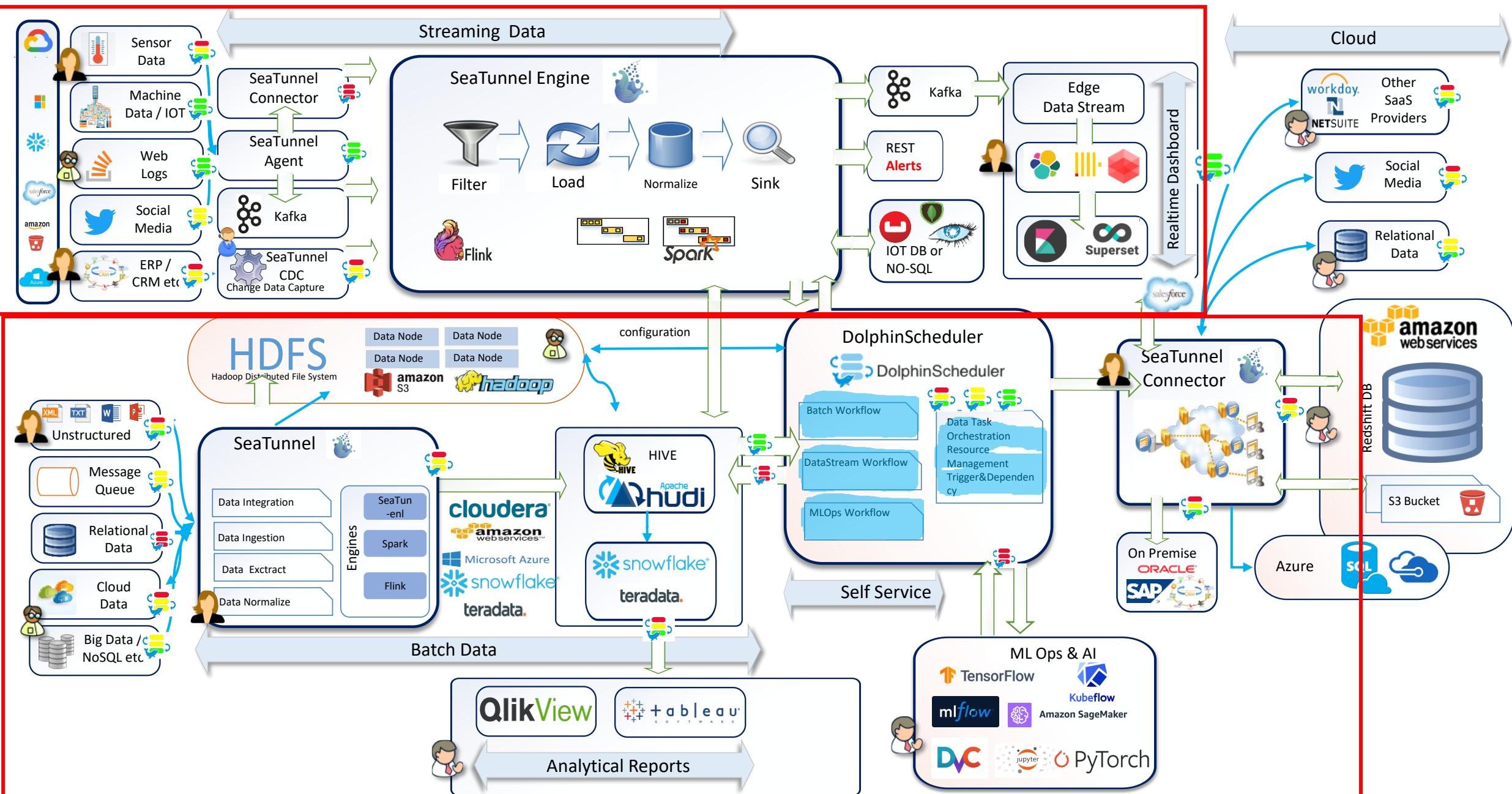
Track Chair of Workflow/Data Governance of Apache Con Asia 2021/2022

William used to be CTO of Analysys and Senior Big Data Director of Lenovo, Wanda, CICC, IBM, and Teradata. He has more than 20+years of experience in big data technology and data management.

# Agenda

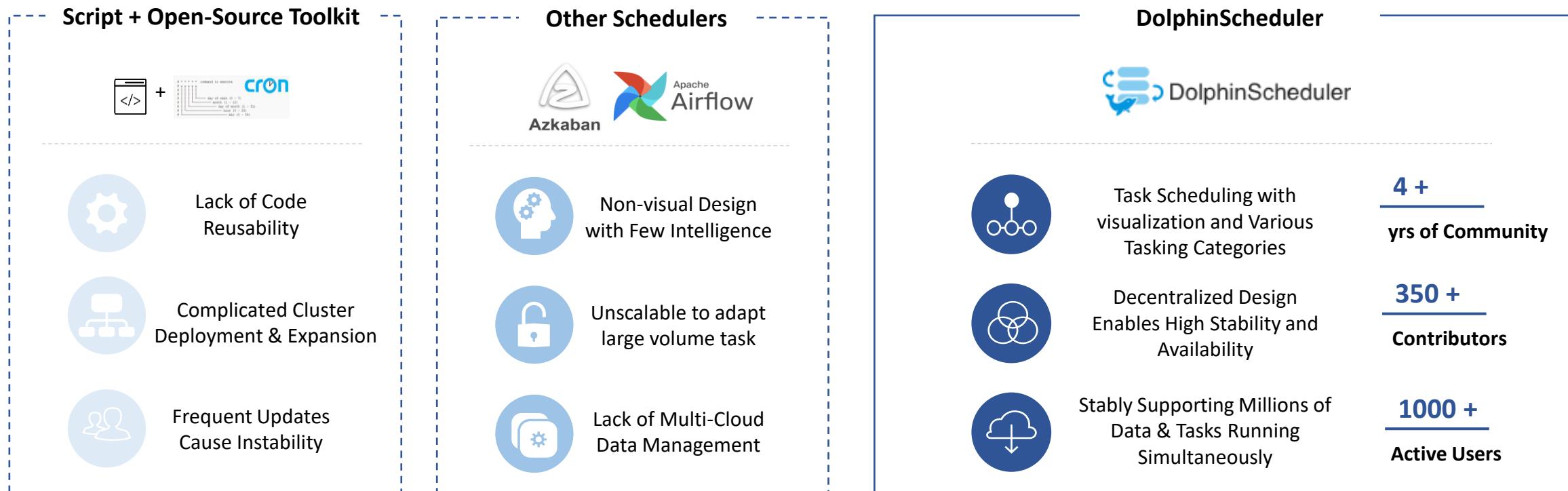
- Introduction of DolphinScheduler
- 2.0 & 3.1.0 New Features
- User Case – Cisco Webx

# Apache Projects in Modern Data Stack & DataOps in Enterprise



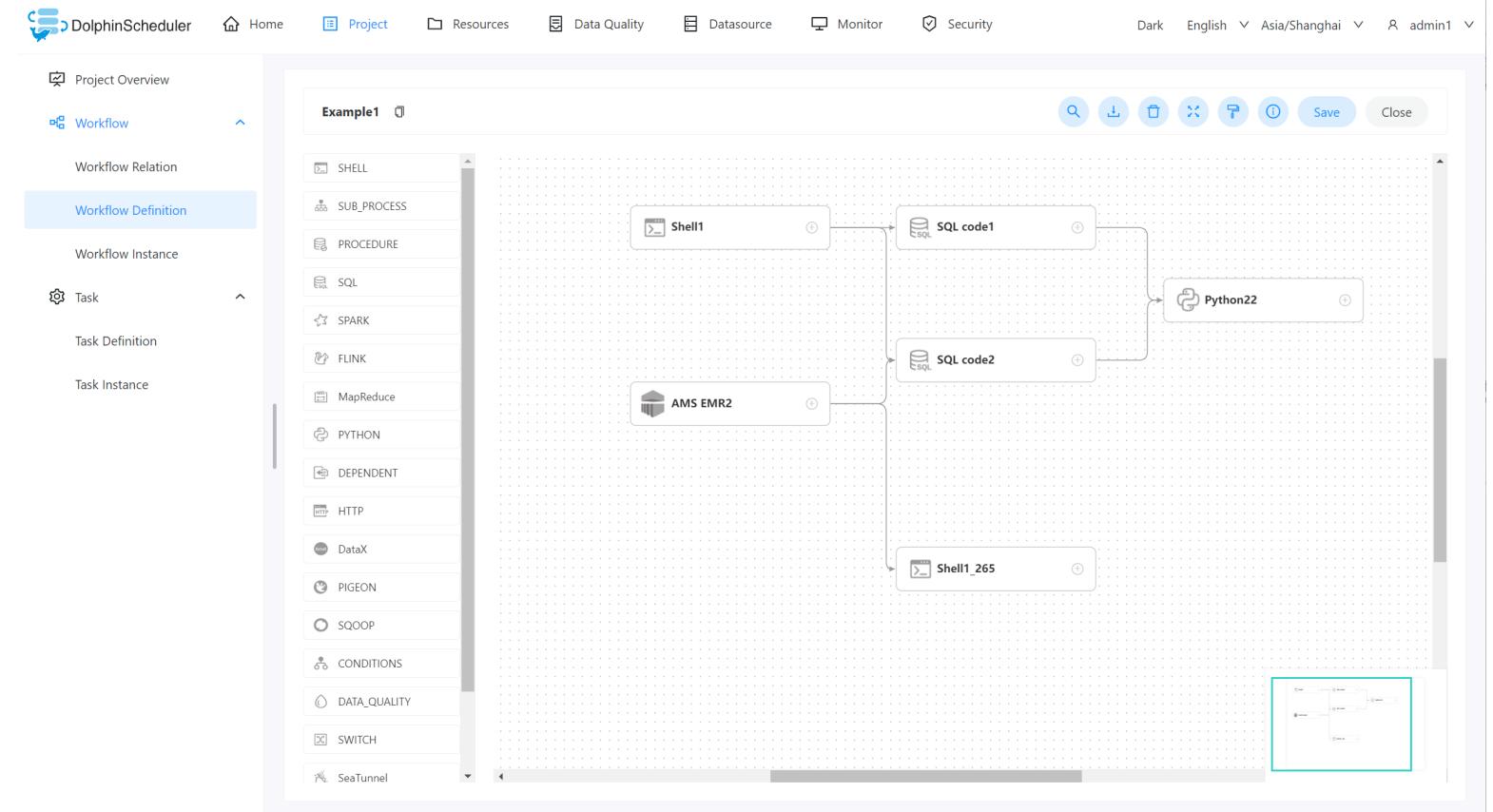
# DolphinScheduler: Cloud-Native Visual Scheduler Engine with High Stability

Star 8.1k

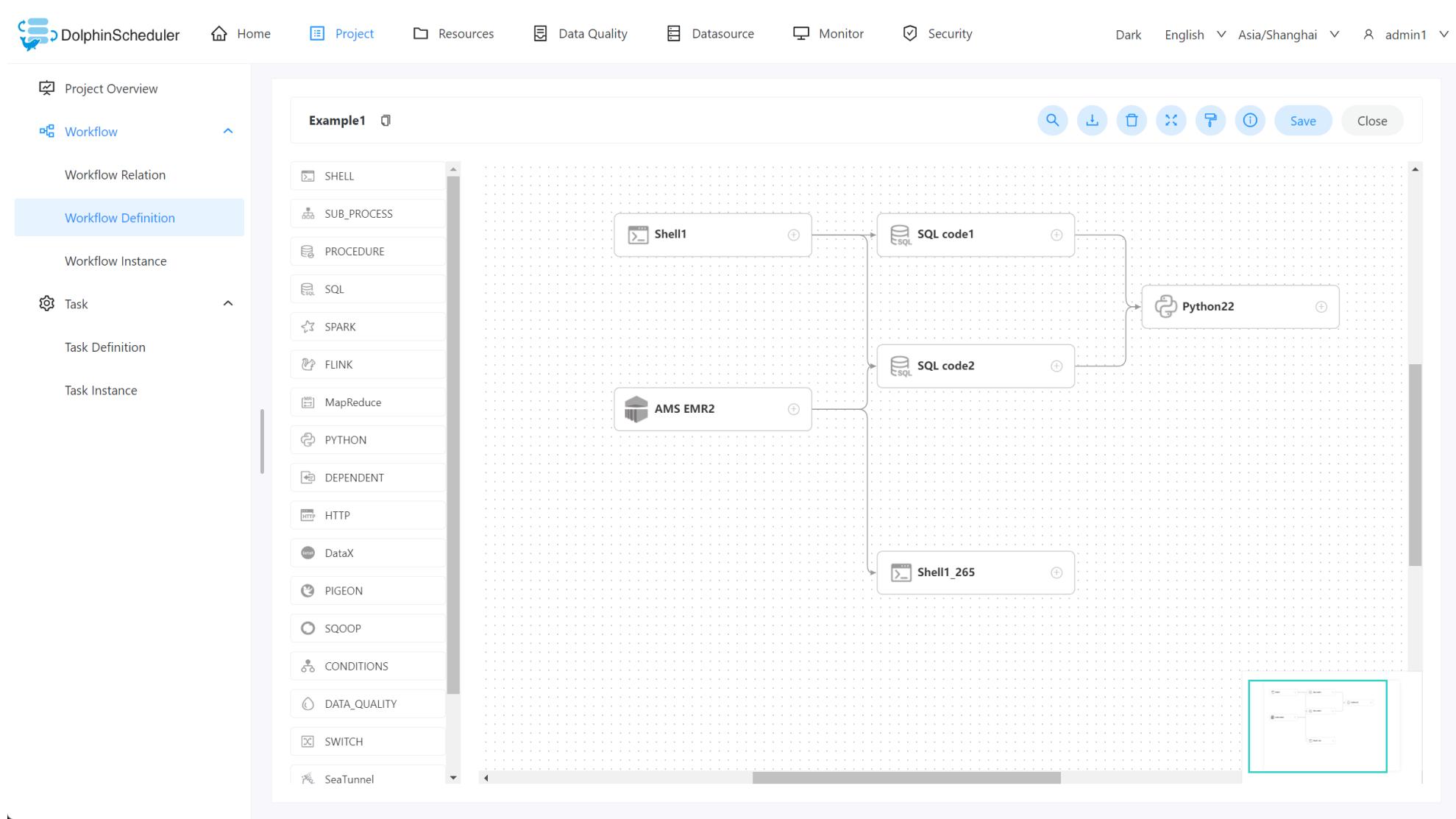


# Apache DolphinScheduler

DolphinScheduler is a distributed and extensible workflow scheduler platform with powerful DAG visual interfaces, dedicated to solving complex task dependencies in the data pipeline and providing various types of jobs.

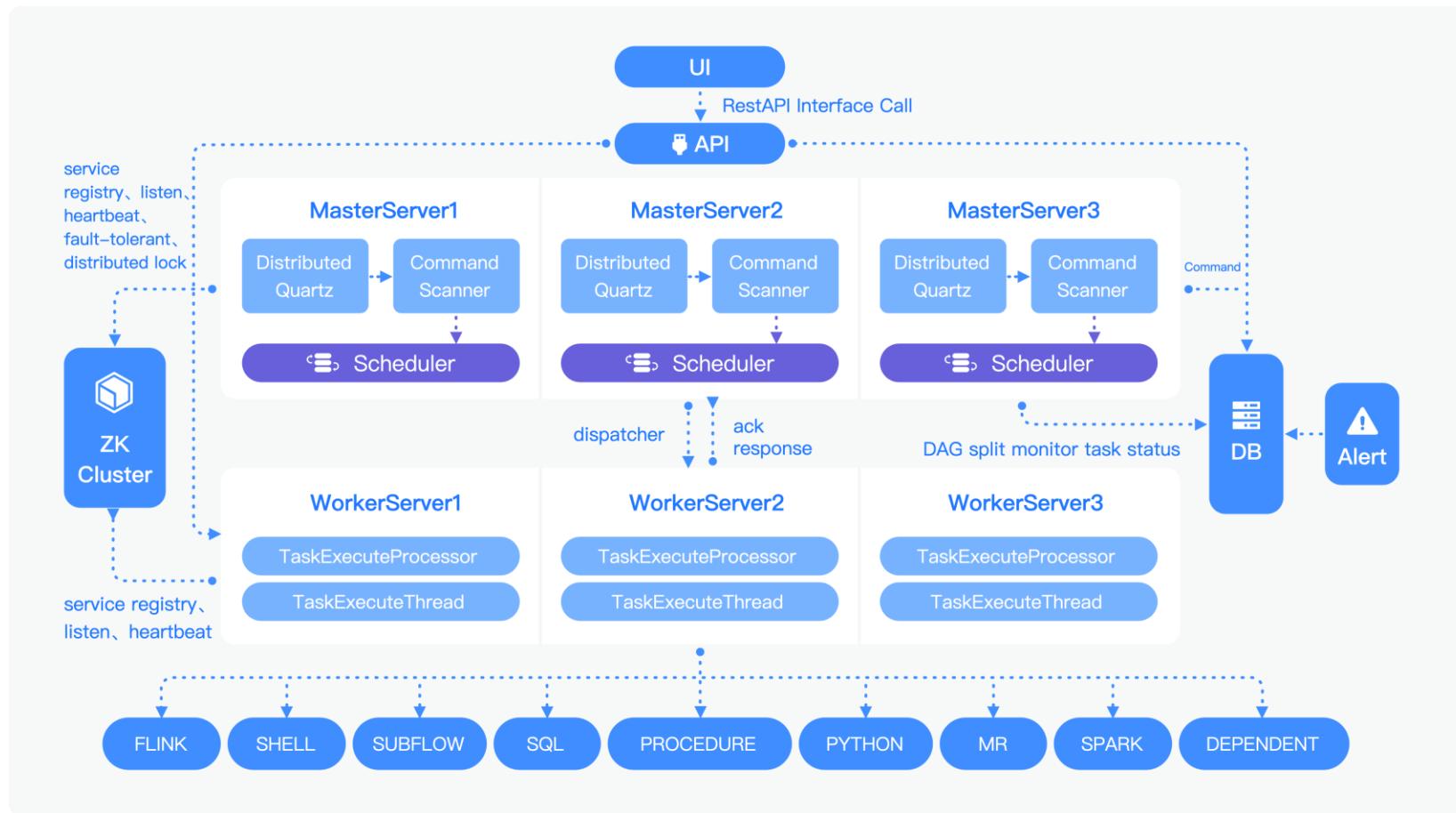


# Drag&Drop To Create a WYSWYG Workflow without Code.

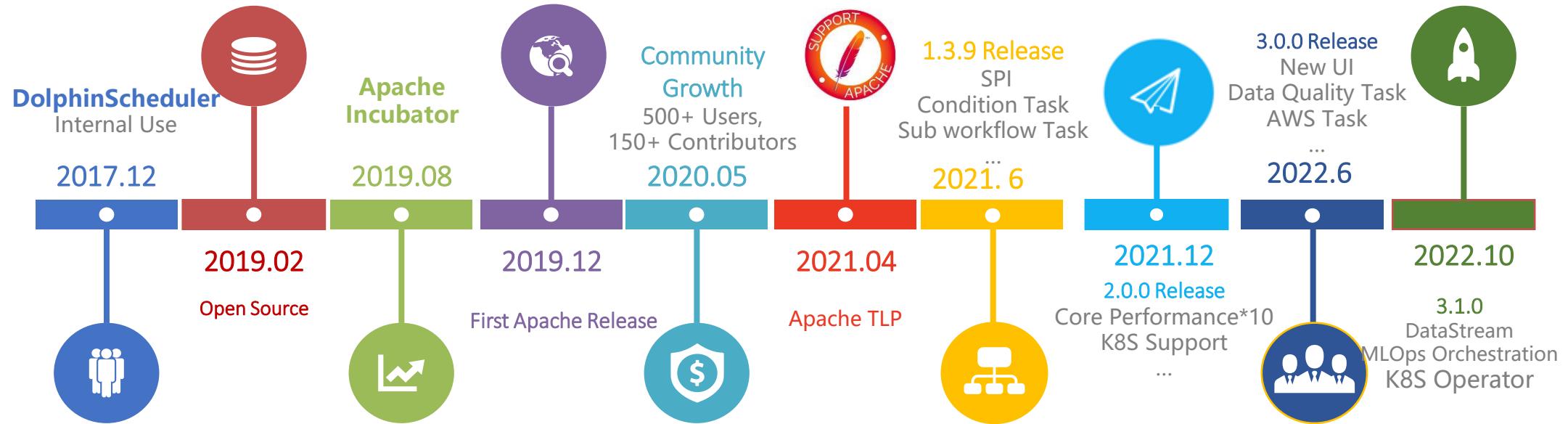


# DolphinScheduler High Performance & High Stability with Multi Workers/Masters

Apache DolphinScheduler is dedicated to solving complex job dependencies in the data pipeline and providing various types of jobs available out of box.



# Apache DolphinScheduler History



# DolphinScheduler Is Trusted by Many Industry Leaders



Tencent



PING AN



inspur

Lenovo

NOKIA

accenture



iFLYTEK

WPS



JD Logistics



BELLE



SheIn



UCLOUD



HUAWEI

vmware

vip.com

JDT



# DolphinScheduler Typical User Case



China Unicom originally used an enterprise scheduling system to support data processing and task scheduling of their global data platform in combination with Shell (HiveSQL) . After comparing Airflow, Azkaban, and other commercial scheduler, China Unicom finally chose DS.

- Met business and scheduling functional needs
- Met large data volume requirements
- Cost-effectiveness



PINGAN

## 1 High-Performance, High-Volume Task Scheduling

# SHEIN

SHEIN originally used Airflow to schedule global tasks; however, Airflow has a centralized design and lack of visualization, and it was also unable to support K8S and globalized cloud-native deployment. Thus, SHEIN chose to migrate from Airflow to DolphinScheduler.

- Global cloud deployment, K8S support
- Decentralization to ensure stability
- Easy to use for data consumers without developer background



## 2 Global Cloud Deployment with Ease of Use for Data Consumers



Litchi FM used SQL/Shell/Python scripts and other big data components for their AI system, which was difficult both to use and to reuse. After using the AI development platform based on DS, Litchi FM abstracted the entire process of from data acquisition to model training and connected them with DAG through DS 's low-code IDE.

- Efficiently computing of massive big data tasks
- Reusable ML process
- DAG execution engine



## 3 AI/ML Orchestration

# Agenda

- Introduction of DolphinScheduler
- 2.0 & 3.1.0 New Features
- User Case – Cisco Webx

# DolphinScheduler 1.x and 2.x Features



- Tasks are associated as DAG form
- Real-time monitoring of task status



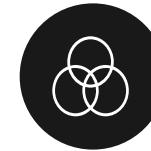
- Supports more than 20 task types such as Shell, MR, Spark, SQL, dependency, etc



- Supports workflow priority, task priority
- Global parameters and **customized** parameters



- Supports workflow scheduling, dependent, manual impacts, and pause/stop/resume



- Supports multi-tenancy, Multi-Projects
- online log viewing and resource online management



- Complete system monitoring, task timeout /failure alert.

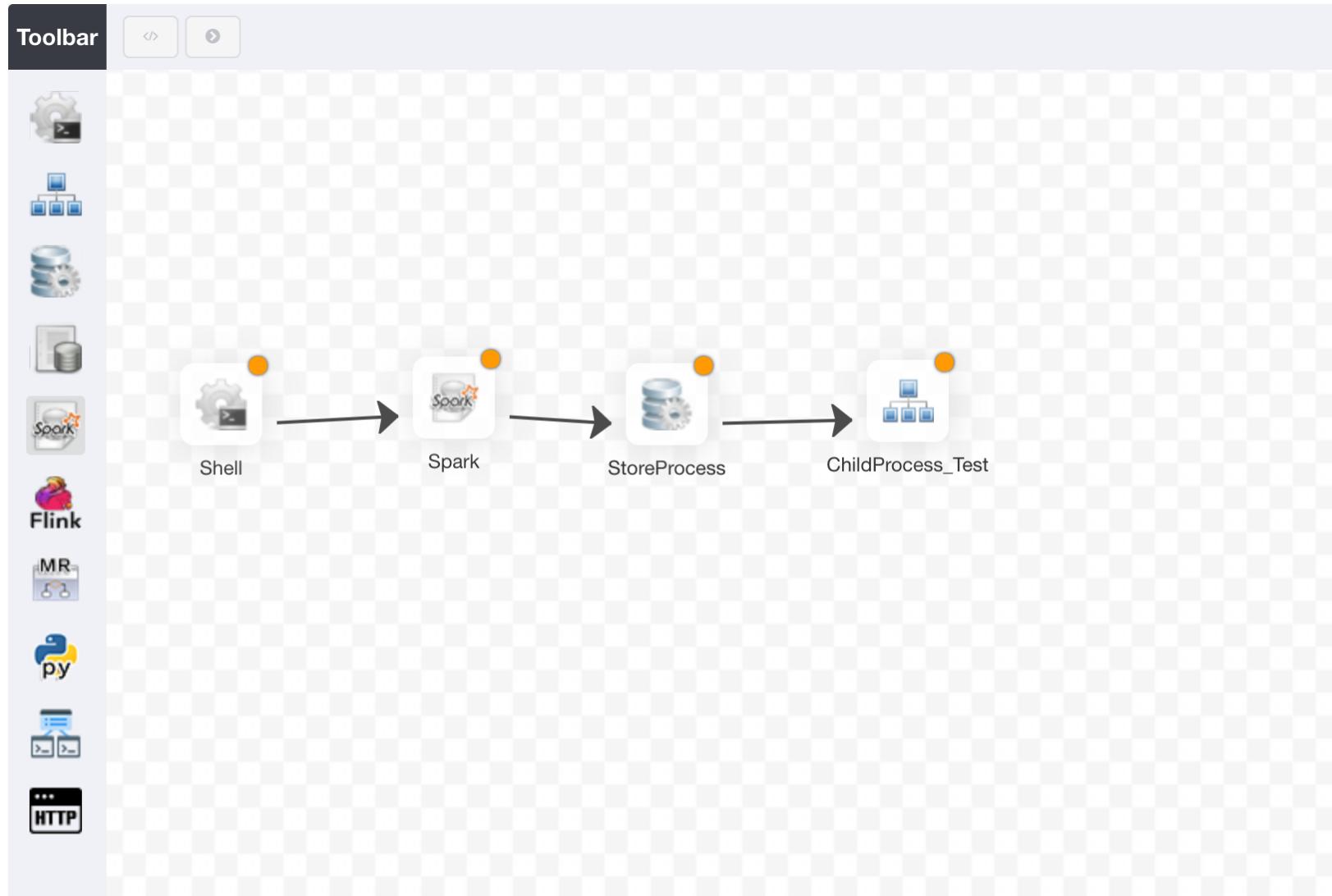


- The decentralized design ensures the stability and high availability of the system



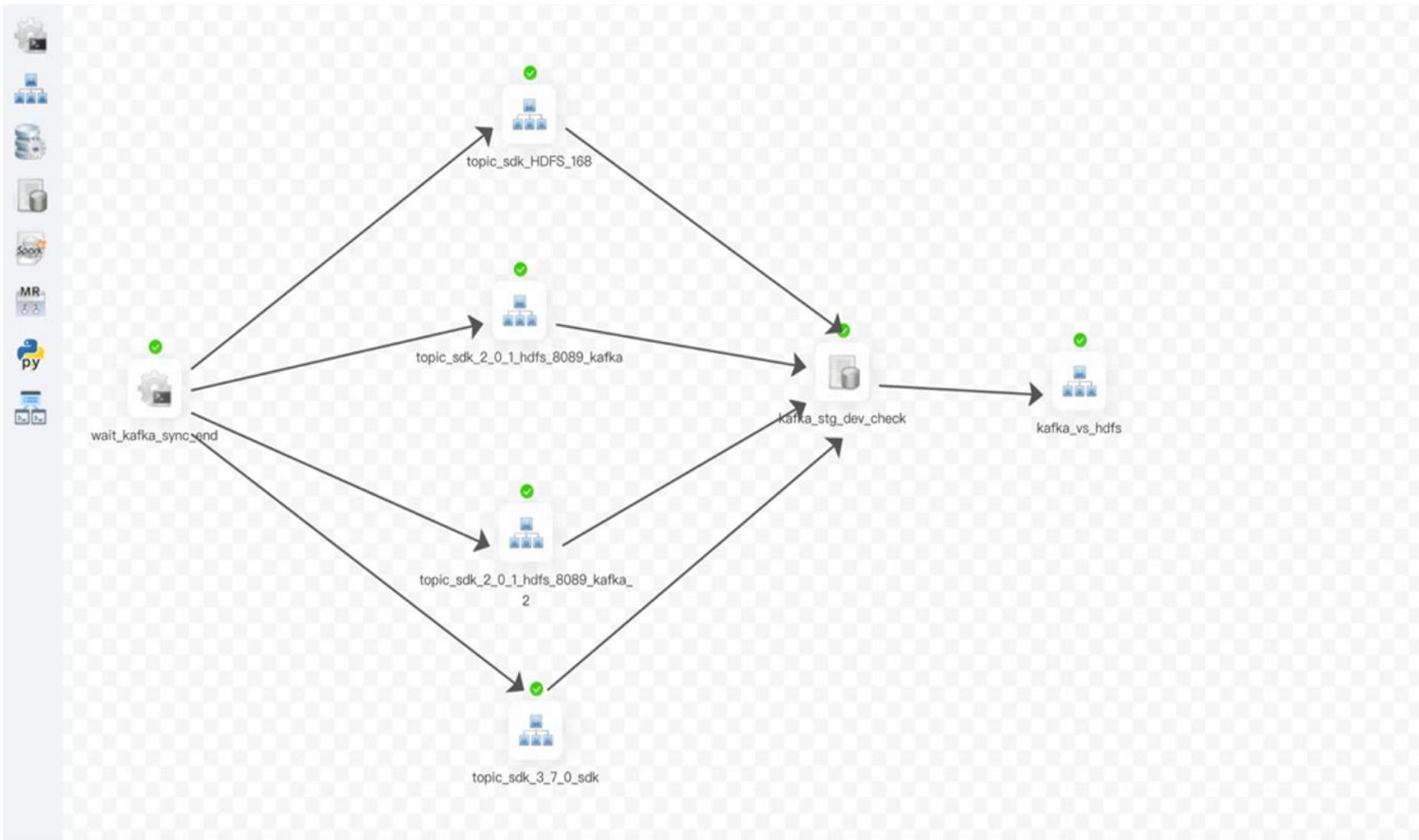
- Supports stable operation of 100,000 volume of data tasks per day

# Workflow Management: Visualized Drag-and-Drop Workflow Configuration



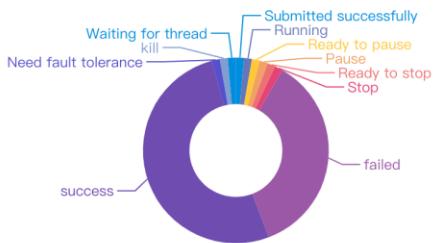
1. Visualized drag-and-drop optimizes task creation efficiency
2. Support various task types: Shell, MR, Spark, SQL (MySQL, PostgreSQL, hive, spark SQL), Python, Sub\_Process, Procedure, etc
3. Sub\_Process
  - Sub\_Process enables the reuse of **data resolve**, imports and data persistence, avoid repeated configurations

# Visualization of Running Workflow



# Task Management: Multi-Level Monitoring

Task Status Statistics



#	Number	State
1	0	Submitted successfully
2	0	Running
3	0	Ready to pause
4	0	Pause
5	0	Ready to stop
6	0	Stop
7	16	failed
8	23	success
9	0	Need fault tolerance
10	0	kill
11	0	Waiting for thread

Task status data statistics

Process Instance

#	Process Name	Run Type	Scheduling Time	Start Time	End Time	Durations	Run Times	host	fault-tolerant sign	StateOperation
1	var_test-0-1600416909502	Start Process	-	2020-09-18 16:15:10	2020-09-18 16:15:14	4	1	192.168.220.241	NO	
2	var_test-0-1599147308051	Start Process	-	2020-09-03 23:35:08	2020-09-03 23:35:13	5	1	192.168.220.241	NO	
3	python_test-0-1597809791894	Start Process	-	2020-08-19 12:03:12	2020-08-19 12:03:16	4	1	192.168.220.241	NO	
4	python_test-0-1597809507340	Start Process	-	2020-08-19 11:58:27	2020-08-19 11:58:31	4	1	192.168.220.241	NO	
5	var_test-0-1596276257320	Start Process	-	2020-08-01 18:04:17	2020-08-01 18:04:21	4	1	192.168.220.241	NO	
6	var_test-0-1593598311471	Start Process	-	2020-07-01 18:11:51	2020-07-01 18:11:55	4	1	192.168.220.241	NO	
7	var_test-0-1593598260252	Scheduling execution	2020-07-01 18:11:00	2020-07-01 18:11:00	2020-07-01 18:11:05	5	1	192.168.220.241	NO	
8	var_test-0-1593598200922	Scheduling execution	2020-07-01 18:10:00	2020-07-01 18:10:01	2020-07-01 18:10:05	4	1	192.168.220.241	NO	
9	var_test-0-1593598140671	Scheduling execution	2020-07-01 18:09:00	2020-07-01 18:09:01	2020-07-01 18:09:05	4	1	192.168.220.241	NO	
10	var_test-0-1593598021027	Scheduling execution	2020-07-01 18:07:00	2020-07-01 18:07:01	2020-07-01 18:07:05	4	1	192.168.220.241	NO	

1 2 3 4 5 6 7 > 10/page Go to

Process instance status view

View log

```
[INFO] 2020-09-18 16:15:09.893 - [taskAppId=TASK-80-2774-11870]:[84] - python task params {"rawScript":"print(1111)","localParams":[],"resourceList":[]}
[INFO] 2020-09-18 16:15:09.894 - [taskAppId=TASK-80-2774-11870]:[131] - raw python script : print(1111)
[INFO] 2020-09-18 16:15:09.894 - [taskAppId=TASK-80-2774-11870]:[132] - task dir : /tmp/dolphinscheduler/exec/process/1/80/2774/11870
[INFO] 2020-09-18 16:15:09.896 - [taskAppId=TASK-80-2774-11870]:[329] - task run command:
sudo -u hdfs /usr/bin/python /tmp/dolphinscheduler/exec/process/1/80/2774/11870/py_80_2774_11870.command
[INFO] 2020-09-18 16:15:09.898 - [taskAppId=TASK-80-2774-11870]:[158] - process start, process id is: 3111
[INFO] 2020-09-18 16:15:09.914 - [taskAppId=TASK-80-2774-11870]:[106] - -> 1111
[INFO] 2020-09-18 16:15:09.916 - [taskAppId=TASK-80-2774-11870]:[168] - process has exited, work dir:/tmp/dolphinscheduler/exec/process/1/80/2774/11870, pid:3111,exitStatusCode:0
[INFO] 2020-09-18 16:15:09.919 - [taskAppId=TASK-80-2774-11870]:[231] - process id is 3111
```

Gantt



Tracking of task execution status

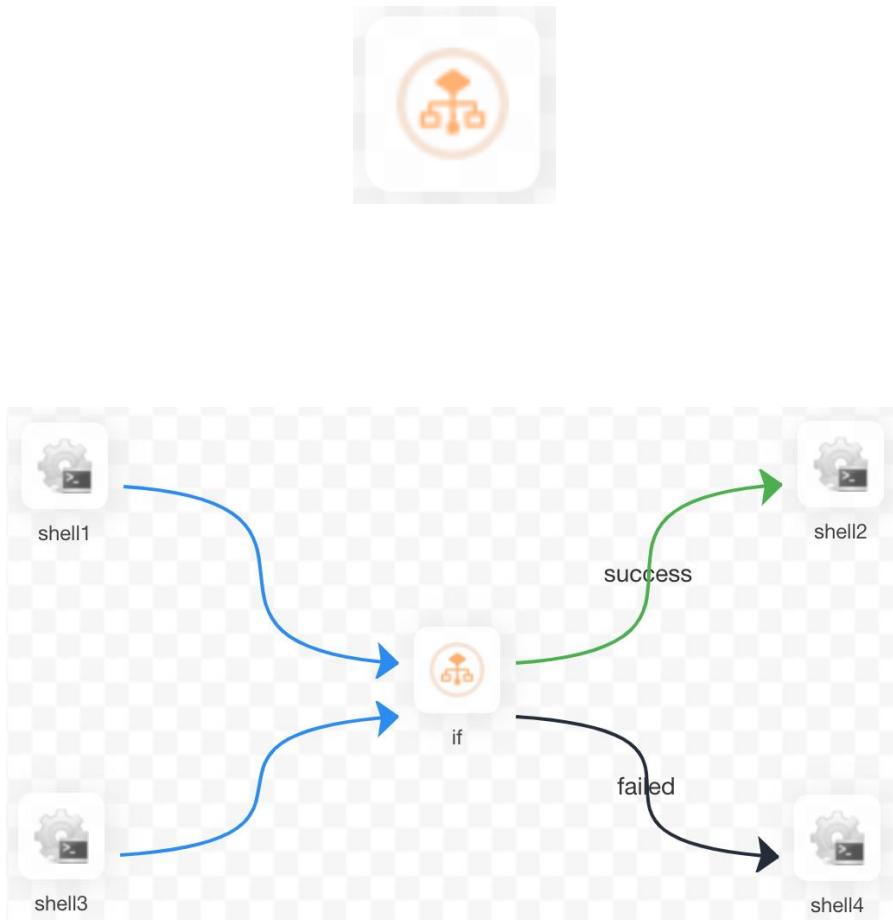
Task execution log online

# Data-Source Management: Visualized Configuration and Multiple Data Compatibility

The screenshot shows the DolphinScheduler web interface. The top navigation bar includes links for Home, Project, Resources, Datasource (which is currently selected), Monitor, and Security. The user is logged in as 'admin'. The main content area has two sections: 'Datasource' and 'Create Datasource'. The 'Datasource' section lists ten existing data sources with names like 'jg2', 'test6', 'jg', '本地', 'zhuizhen', 'zakimysql', 'gp', '12', 'all\_mysql', and 'PG数据源'. The 'Create Datasource' section is a form titled 'CreateDatasource' with fields for: Datasource Type (MySQL is selected), Datasource Name (input field 'Please enter datasource name'), Description (input field 'Please enter description'), IP (input field 'Please enter IP'), Port (input field '3306'), User Name (input field 'Please enter user name'), Password (input field 'Please enter your password'), Database Name (input field 'Please enter database name'), and JDBC connect parameters (input field 'Please enter format {"key1":"value1","key2":"value2"} connection parameter'). It also features 'Test Connect' and 'Submit' buttons. To the right of the form is a table listing data source details such as Create Time, Update Time, and Operation (with edit and delete icons). A search bar at the top right allows filtering by keyword.

1. Visualized data-sources include : MySql、PostgerSql、Hive、Impala、Spark、ClickHouse、Oracle、SqlServer、DB2、MongoDB.
2. Supports Plugin data-source extension
3. Visualized data-source management
4. Configure once, use everywhere.

# DolphinScheduler Condition Task



Current node settings

Node name

Run flag  Normal  Prohibition execution

Description

Task priority

Worker group

Number of failed retries  (Times) Failed retry interval  (Minute)

State

Branch flow

State

Branch flow

Timeout alarm   
Custom Parameters

# Multi-Cloud, Multi-K8s, Muti Big Data Environment Support

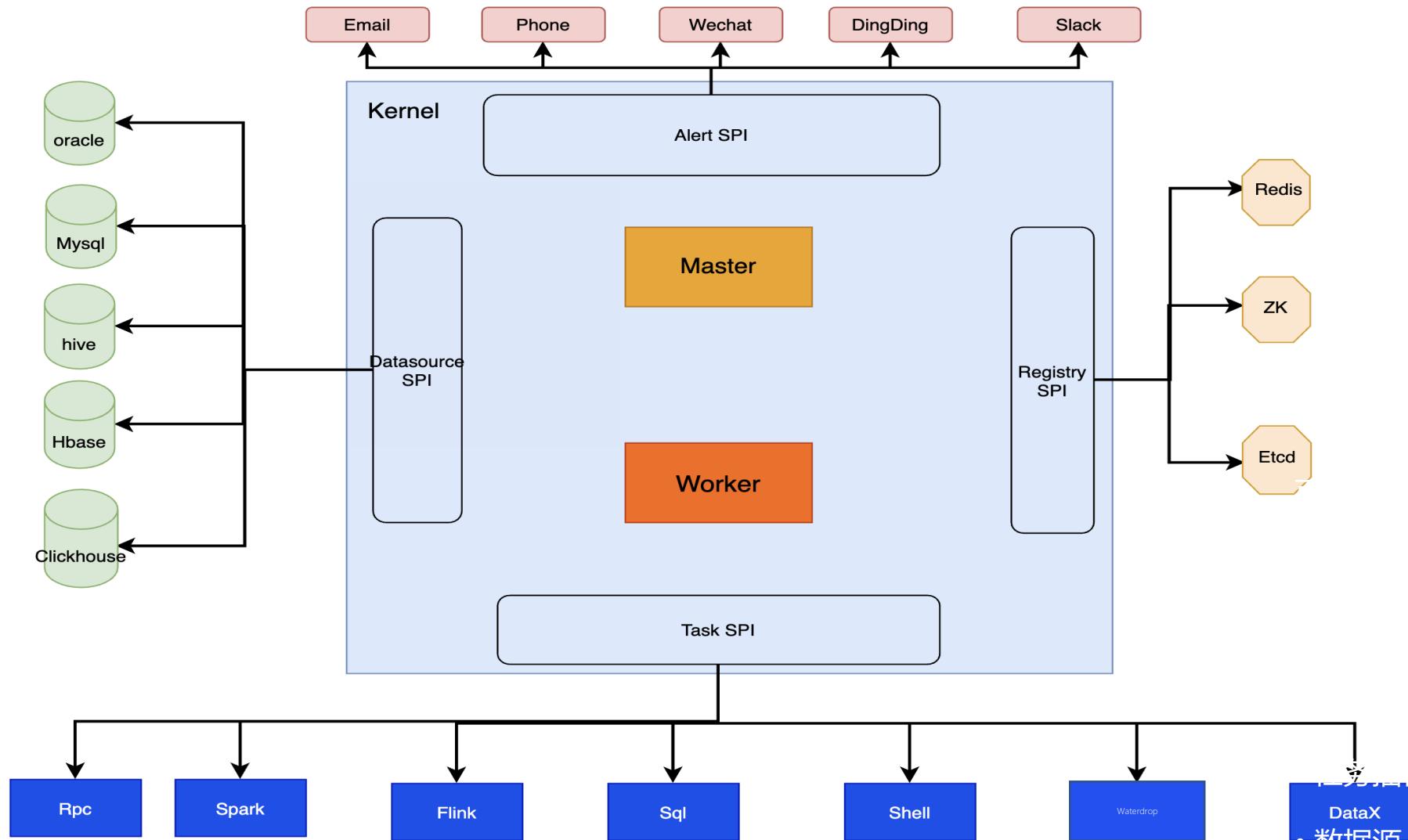
- Different Environment Configuration
- Task mapping with Environment
- Different WorkerGroup Environment Configuration

Create Environment X

Name	env-hadoop
Configuration	<pre>export HADOOP_HOME=/opt/soft/hadoop export HADOOP_CONF_DIR=/opt/soft/hadoop/etc/ hadoop export SPARK_HOME1=/opt/soft/spark1 export SPARK_HOME2=/opt/soft/spark2 export PYTHON_HOME=/opt/soft/python export JAVA_HOME=/opt/soft/java export HIVE_HOME=/opt/soft/hive export FLINK_HOME=/opt/soft/flink export DATAx_HOME=/opt/soft/datax  export PATH=\$HADOOP_HOME/bin:\$SPARK_HOM E1/bin:\$SPARK_HOME2/bin:\$PYTHON_HOME/bi n:\$JAVA_HOME/bin:\$HIVE_HOME/bin:\$FLINK_HO ME/bin:\$DATAx_HOME/bin:\$PATH</pre>
Description	hadoop-env
Worker Group	default <span style="border: 1px solid #ccc; padding: 2px;">X</span>

取消 提交

# Service Provide Interface – Easy to Extend Your Own Task & DataSource



# DolphinScheduler 3.1.0 New Features

2.X  
version



3.1.0  
New  
Feature



## Simple& WYSWYG workflow

- Drag & Drop to create workflow
- DAG Graph run-time management
- Open API to support others



## High Reliability

- Decentralized multi-Masters and multi-Worker
- High performance(support 1m+ Task in production env)
- High Reliability



## Rich Workflow Functions

- Support pause&resume workflow
- Support projects,multi-tenant
- Support 30+ Tasktype, Spark, Hive, MR, Python, Sub-Process, Shell,EMR, S3



## Cloudnative&Extensible

- Support User-defined Task
- Condition and subworkflow
- Elastic Master & Worker dynamic on-line&off-line



## ML Orchestration

- DataPreparation+MLOps
- ML flow, Sagemaker,DVC
- Jupyter, PyTorch
- Kubeflow, TensorFlow,Bentoml...



## Data Stream Support

- Flink, Sparking streaming Support
- Data Stream workflow Support
- Data Stream Management



## Python, YAML Workflow Support

- Python generate Workflow
- YAML generate Workflow
- Code Review & Deployment



## K8S Support

- K8S Operator
- K8S Task

# New Feature PyDolphinScheduler

PyDolphinScheduler is Python API for Apache DolphinScheduler, which allow you definition your workflow by Python code, aka workflow-as-codes.

Python

```
# [start package_import]
# Import ProcessDefinition object to define your workflow attributes
from pydolphinscheduler.core.process_definition import ProcessDefinition

# Import task Shell object cause we would create some shell tasks later
from pydolphinscheduler.tasks.shell import Shell

# [end package_import]

# [start workflow_declare]
with ProcessDefinition(
    name="tutorial",
    schedule="0 0 0 * * ? *",
    start_time="2021-01-01",
    tenant="tenant_exists",
) as pd:
    # [end workflow_declare]
    # [start task_declare]
    task_parent = Shell(name="task_parent", command="echo hello pydolphinscheduler")
    task_child_one = Shell(name="task_child_one", command="echo 'child one'")
    task_child_two = Shell(name="task_child_two", command="echo 'child two'")
    task_union = Shell(name="task_union", command="echo union")
    # [end task_declare]

    # [start task_relation_declare]
    task_group = [task_child_one, task_child_two]
    task_parent.set_downstream(task_group)

    task_union << task_group
    # [end task_relation_declare]

    # [start submit_or_run]
    pd.run()
    # [end submit_or_run]
```

YAML

```
# Define the process
process:
    name: "tutorial"
    schedule: "0 0 0 * * ? *"
    start_time: "2021-01-01"
    tenant: "tenant_exists"
    release_state: "offline"
    run: true

# Define the tasks under the process
tasks:
    -
        task_type: Shell
        params:
            name: task_parent
            command: |
                echo hello pydolphinscheduler

    -
        task_type: Shell
        deps: [task_parent]
        params:
            name: task_child_one
            command: echo "child one"

    -
        task_type: Shell
        deps: [task_parent]
        params:
            name: task_child_two
            command: echo "child two"

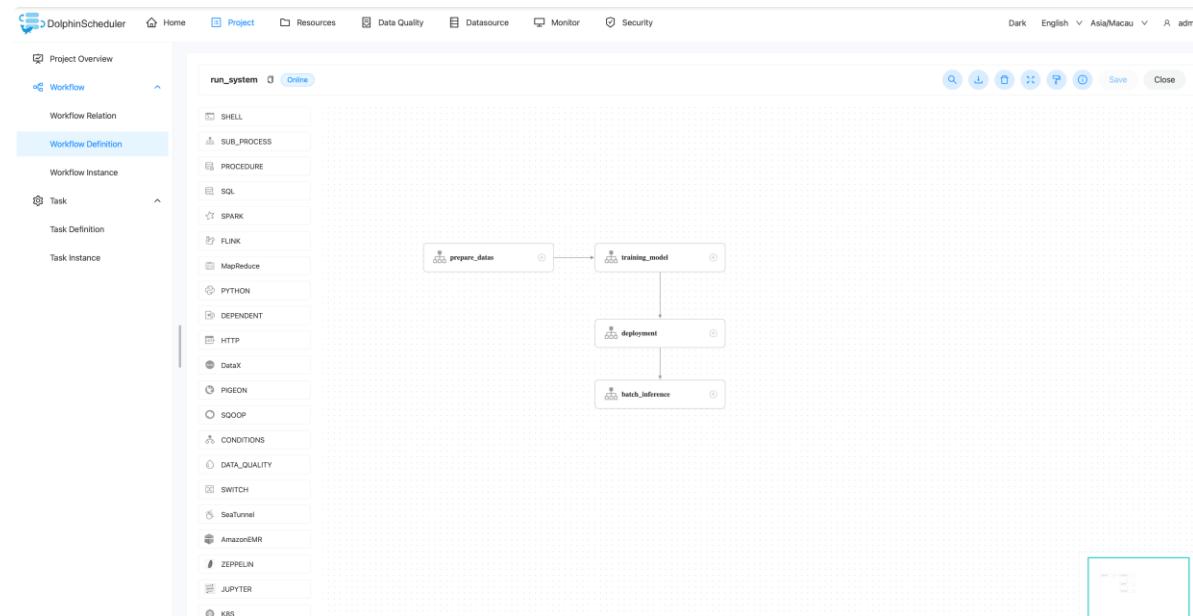
    -
        task_type: Shell
        deps: [task_child_one, task_child_two]
        params:
            name: task_union
            command: echo "union"
```

# New Feature DolphinScheduler ML Orchestration x MLOps

In the field of MLOps, DolphinScheduler is adding a variety of machine learning-related task plugin to help data analysts and data scientists easily use DolphinScheduler. Solve the following two problems:

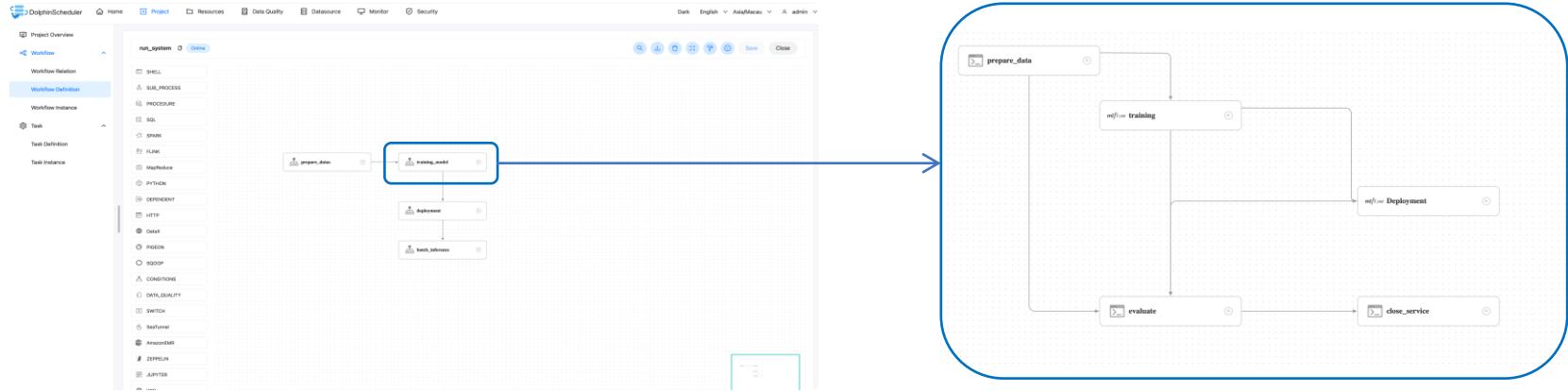
**The efficiency of the machine learning lifecycle**

**The efficiency of machine learning systems to connect with other systems**

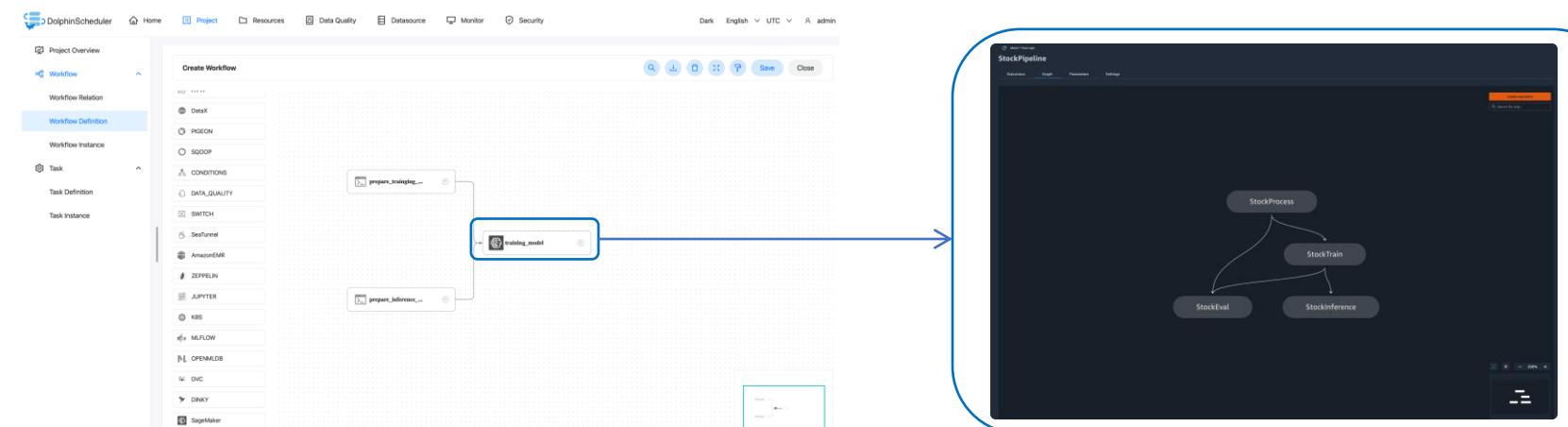


# MLOps Orchestration

## Machine learning workflows in DolphinScheduler



## Machine learning workflows between Spark and SageMaker



# MLOps Orchestration

## Task plugins support for machine learning workflow

Data Management	Feature Store	Model Training	Deployment	Model Management
DVC, SageMaker	OpenMLDB, SageMaker	Shell, Python, Jupyter, MLflow, Pytorch, SageMaker	Shell, Python, MLflow, SageMaker	MLflow, SageMaker

## Task plugins in DolphinScheduler MLOps Orchestration

Task Plugin	Scenario
Jupyter	Schedule the execution of model training, data analysis notebook Add the notebook to the workflow
MLflow	Run the custom MLFlow Project, built-in algorithms, AutoML Deploy machine learning models
OpenMLDB	Feature extraction and calculation for offline and online consistency
DVC	Upload and download data based on version information Large file version management based on Git repository
SageMaker	Schedule the execution of SageMaker Pipeline Connect tasks such as upstream big data analytics or some downstream tasks
Pytorch	Migrate the machine learning project to DolphinScheduler Run a Git-based machine learning project Also can run Tensorflow and other ML project

# Jupyter Task Plugin

Jupyter task plugin can create a jupyter-type task and execute jupyter notes, it will use papermill to evaluate jupyter notebooks.

The form consists of several input fields:

- condaEnvName \*: Please enter the conda environment name of papermill
- inputNotePath \*: Please enter the input jupyter note path
- outputNotePath \*: Please enter the output jupyter note path
- parameters: Please enter the parameters for jupyter parameterization
- kernel: Please enter the jupyter kernel name
- engine: Please enter the engine name
- executionTimeout: Please enter the execution timeout for each jupyter note cell
- startTimeout: Please enter the start timeout for jupyter kernel
- others: Please enter the other options you need for papermill
- Resources: Please select resources
- Custom Parameters: A blue circular icon.

## Scenario:

1. Schedule to execute machine learning notebooks such as model training
2. Schedule to execute data analysis and data visualization notebooks
3. Run the notebook with different parameters

# MLflow Task Plugin

MLflow task plugin used to execute MLflow tasks, Currently contains MLflow Projects and MLflow Models.

MLflow Tracking Server URI	MLflow Task Type
http://127.0.0.1:5000	MLflow Projects
Job Type	Experiment Name
Custom Project	experiment_001
Parameters	
-P learning_rate=0.2 -P colsample-bytree=0.8 -P subsample=0.9	
Repository	
https://github.com/mlflow/mlflow#examples/xgboost/xgboost_native	
Project Version	
master	
Pre tasks	
Please Select	

## Scenario:

1. Run the preset algorithm
2. Run custom MLflow Project
3. Deploy the MLFlow model

# OpenMLDB Task Plugin

OpenMLDB task plugin used to execute tasks on OpenMLDB cluster, provide FeatureStore capability

zookeeper address \*

zookeeper path \*

Execute Mode

 offline  online

SQL Statement \*

```
1 select is_attributed, ip, app, device, os, channel, ho
2 count(channel) over w1 as qty,
3 count(channel) over w2 as ip_app_count,
4 count(channel) over w3 as ip_app_os_count
5 from demo_db.talkingdata
6 window
7 w1 as (partition by ip order by click_time ROWS RANGE
8 w2 as(partition by ip, app order by click_time ROWS R
9 w3 as(partition by ip, app, os order by click_time ROV
10 INTO OUTFILE '/tmp/train_feature';
```

Scenario:

1. Offline feature extraction
2. Online feature extraction
3. Online and offline consistency

# DVC Task Plugin

DVC task plugin is used to use the data version management function of DVC on DolphinScheduler, helping users to carry out data version management easily.

DVC Task Type  
Upload

DVC Repository \*  
git@github.com:<YOUR-NAME-OR-ORG>/dev-data-repository-example.git

Data Path in DVC Repository \*  
iris

Data Path In Worker \*  
/home/ubuntu/data/iris

Version \*  
iris\_20220830

Version Message \*  
ini iris data

Pre tasks  
Please Select

Scenario:

1. Upload the data and record the version
2. Download version-specific data
3. Large file version management based on Git repository

# Amazon SageMaker Task Plugin

SageMaker task plugin can start a SageMaker pipeline execution and use DolphinScheduler to connect other upstream and downstream tasks.

Task priority \*

MEDIUM

Worker group \*

default

Environment Name

Please Select

Task group name

Please Select

Priority

Please Input

Number of failed retries

0 Times - +

Failed retry interval

1 Minute - +

Delay execution time

0 Minute - +

Timeout alarm

SagemakerRequestJson \*

```
1 {  
2     "ParallelismConfiguration":{  
3         "MaxParallelExecutionSteps":1  
4     },  
5     "PipelineExecutionDescription":"test Pipeline",  
6     "PipelineExecutionDisplayName":"AbalonePipeline",  
7     "PipelineName":"AbalonePipeline",  
8     "PipelineParameters":  
9     {  
10        {  
11            "Name":"ProcessingInstanceType",  
12            "Value":"ml.m4.xlarge"  
13        },  
14        {  
15            "Name":"ProcessingInstanceCount",  
16            "Value":"2"  
17        }  
18    }  
19}
```

Custom Parameters

## Scenario:

1. Schedule the execution of SageMaker Pipeline
2. Connect tasks such as upstream big data analytics or some downstream tasks

# Pytorch Task Plugin

Pytorch task plugin enables users to run Pytorch projects in DolphinScheduler more conveniently. In addition, it supports handy Python environment management.

Python Script  
main.py

Script Input Parameters  
--dry-run --no-cuda

Show More Configurations

Project Path  
https://github.com/pytorch/examples.git#mnist

Create An Environment Or Not

Python Environment Manager Tool  
conda

Requirement File  
requirements.txt

Python Version  
3.7

Resources  
Please select resources

Custom Parameters

## Scenario:

1. Migrate the machine learning project to DolphinScheduler
2. Run a Git-based machine learning project
3. Also can run Tensorflow and other ML project

# MLOps Orchestration

Current and future supported machine learning projects.

The current support



Amazon SageMaker



The future support



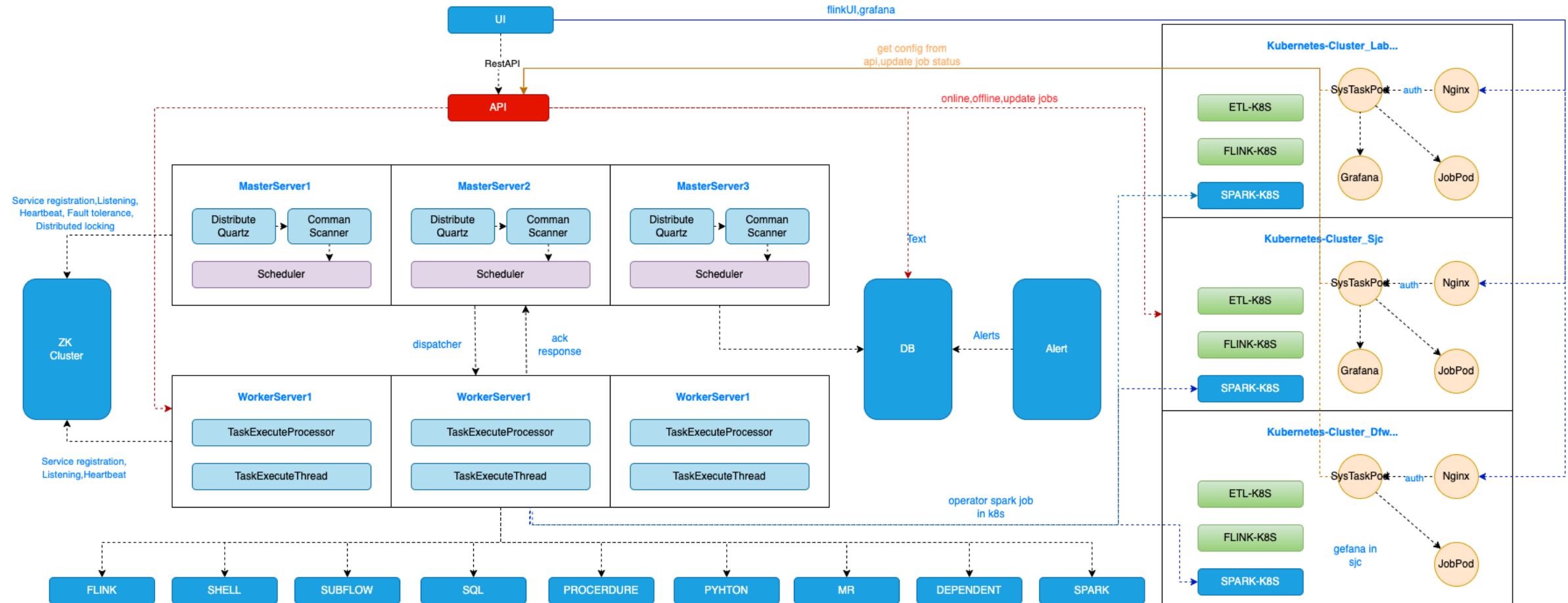
TensorFlow



# Agenda

- Introduction of DolphinScheduler
- 2.0 & 3.1.0 New Features
- User Case – Cisco Webx

# DolphinScheduler with Kubernetes Integration



# Kubernetes Multi-Cluster Management

- Import Kubernetes cluster by input cluster config on the portal
- Support both self-built Kubernetes cluster and public cloud managed Kubernetes cluster, ex Elastic Kubernetes Service

## Create Cluster

Cluster Name \*

Please enter your cluster name

Please enter your cluster name

Kubernetes Config

```
apiVersion: v1
clusters:
- cluster:
  certificate-authority-data: LS0tLS1CZJQ0FURS0tLS0tCg==
  server: https://127.0.0.1:6443
  name: kubernetes
contexts:
- context:
  cluster: kubernetes
  user: kubernetes-admin
  name: kubernetes-admin@kubernetes
current-context: kubernetes-admin@kubernetes
kind: Config
preferences: {}
users:
- name: kubernetes-admin
```

# Kubernetes Multi-Cluster Namespace Management

- Set CPU/Memory Limit for each namespace on different Kubernetes clusters
- Separate resource pools for multi tenancy

The screenshot shows a 'Kubernetes namespace management' interface. On the left, a table lists namespaces across different clusters:

#	Namespace	Cluster
1	[redacted]	bts
2	[redacted]	sjc
3	[redacted]	fra
4	[redacted]	ams
5	[redacted]	fra
6	[redacted]	sjc
7	[redacted]	sjc
8	[redacted]	bts
9		ams
10	flink	bts

A modal window titled 'Edit namespace' is open for the namespace 'flink'. It contains the following fields:

- \* Namespace: flink
- \* Cluster: bts
- Limit CPU: 600 (Core)
- Limit Memory: 4096 (GB)
- Owner: pda

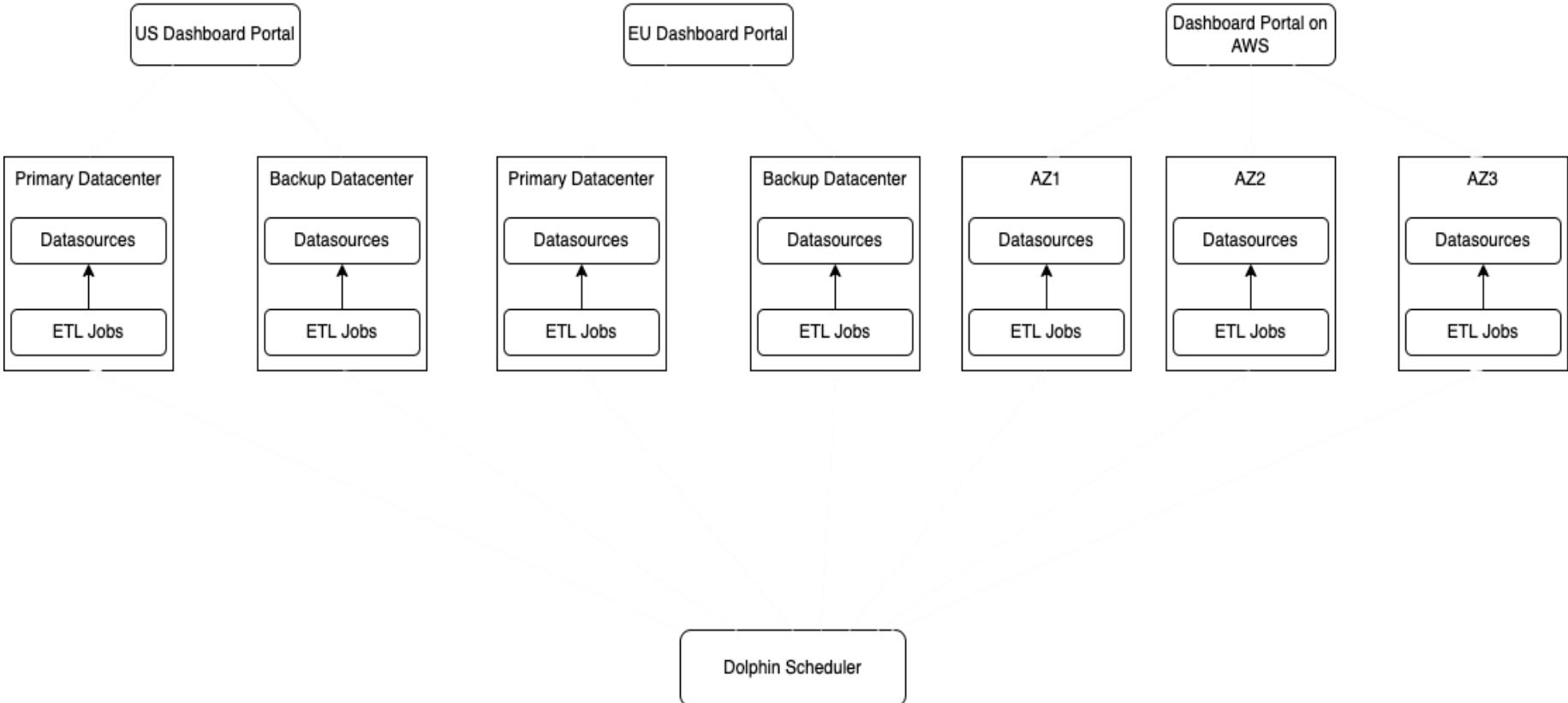
At the bottom of the modal are 'Cancel' and 'Edit' buttons. The background shows a list of namespaces with their last updated time and operation status.

Name	Update Time	Operation
[redacted]	2022-07-04 10:44:53	[edit, delete]
[redacted]	2022-07-01 18:20:49	[edit, delete]
[redacted]	2022-07-01 11:12:52	[edit, delete]
[redacted]	2022-07-01 17:04:29	[edit, delete]
[redacted]	2022-07-01 17:01:04	[edit, delete]
[redacted]	2022-06-30 16:09:38	[edit, delete]
[redacted]	2022-06-27 14:02:57	[edit, delete]
[redacted]	2022-06-27 13:57:54	[edit, delete]
[redacted]	2022-06-24 09:41:43	[edit, delete]
flink	2022-06-23 16:05:40	[edit, delete]

Pagination controls at the bottom include '10/page', page numbers 1-6, and a 'Go to' input field set to '1'.

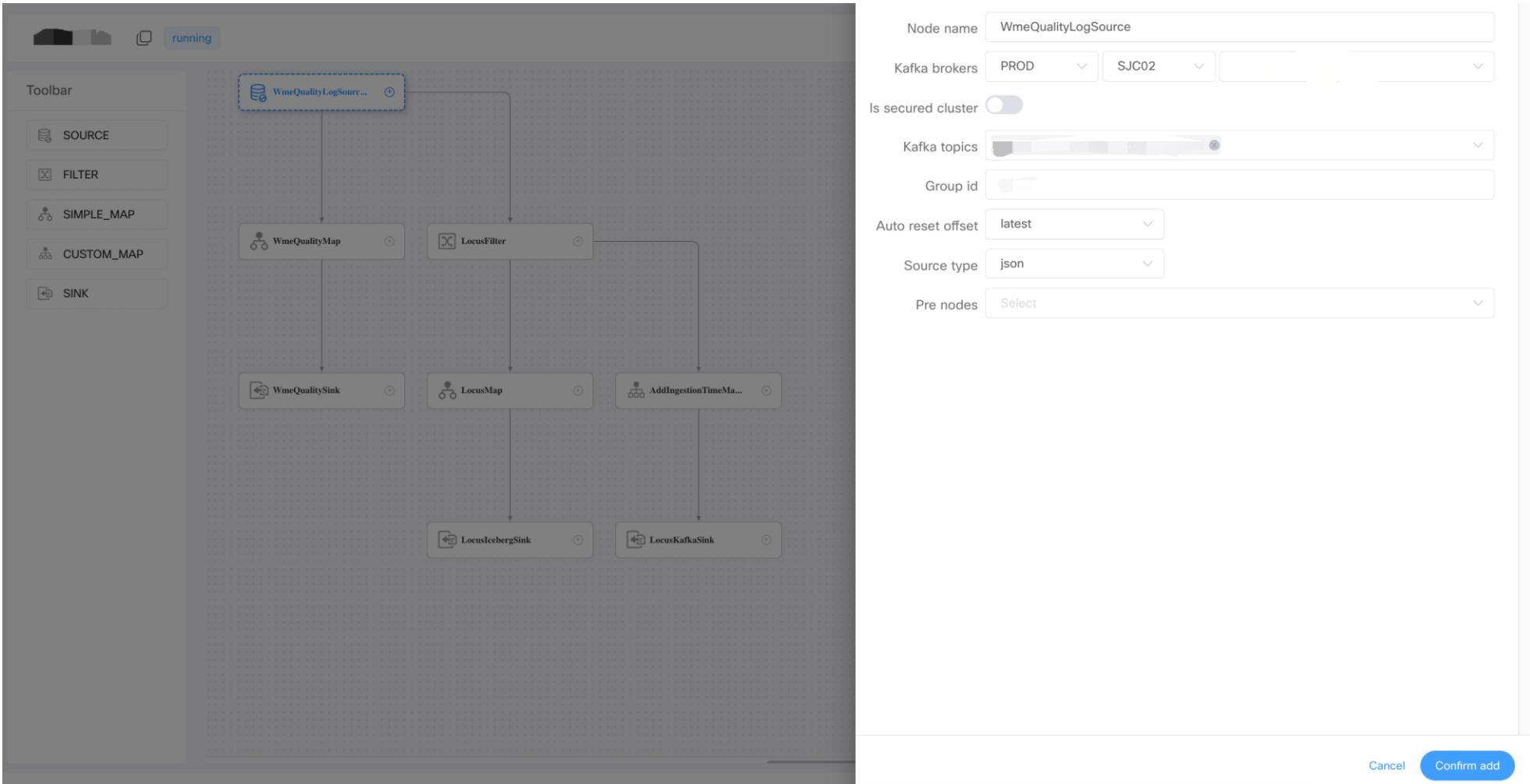
# Multi-Cluster ETL Job Management

- Centralized job scheduling for multiple datacenters across the world
- Both private datacenter and public cloud support



# Simple ETL pipeline

- Generate a complex pipeline by Drag and Drop - No coding required
- Automatic scaling
- Stateless
- UDF support
- Job version management



APACHECON

# Simple ETL pipeline – UDF Management

Node name: QualityMap

Mapping list: Input to research topics

Type	Json Path	Field Name	UDF	Operation
string	report.intervalMetadata.peripherals	report_intervalMetadata_peripherals	select UDF	
string	report.intervalMetadata.periph	UDF Function Name: piiDecrypt Class Name: com.cisco.pda.udf.Decrypt Type: HIVE Jar Package: /common-udf-1.4-snapshot.jar Description: pii Decrypt Function, 0 args	piiHash	
string	report.intervalMetadata.periph	piiDecrypt	piiDecrypt	
string	report.intervalMetadata.periph	piiEncrypt	piiEncrypt	
string	report.intervalMetadata.speakerInfo.info	report_intervalMetadata_speakerInfo_in	getWmeCamera	
string	report.intervalMetadata.microphoneInfo	report_intervalMetadata_microphoneInfo	getWmeMicrophone	
string	report.intervalMetadata.cameralInfo.info	report_intervalMetadata_cameralInfo_in	getWmeSpeaker	
string	reportId	reportId	formatTimestamp	
string	reportVersion	reportVersion	select UDF	
string	reportType	reportType	select UDF	

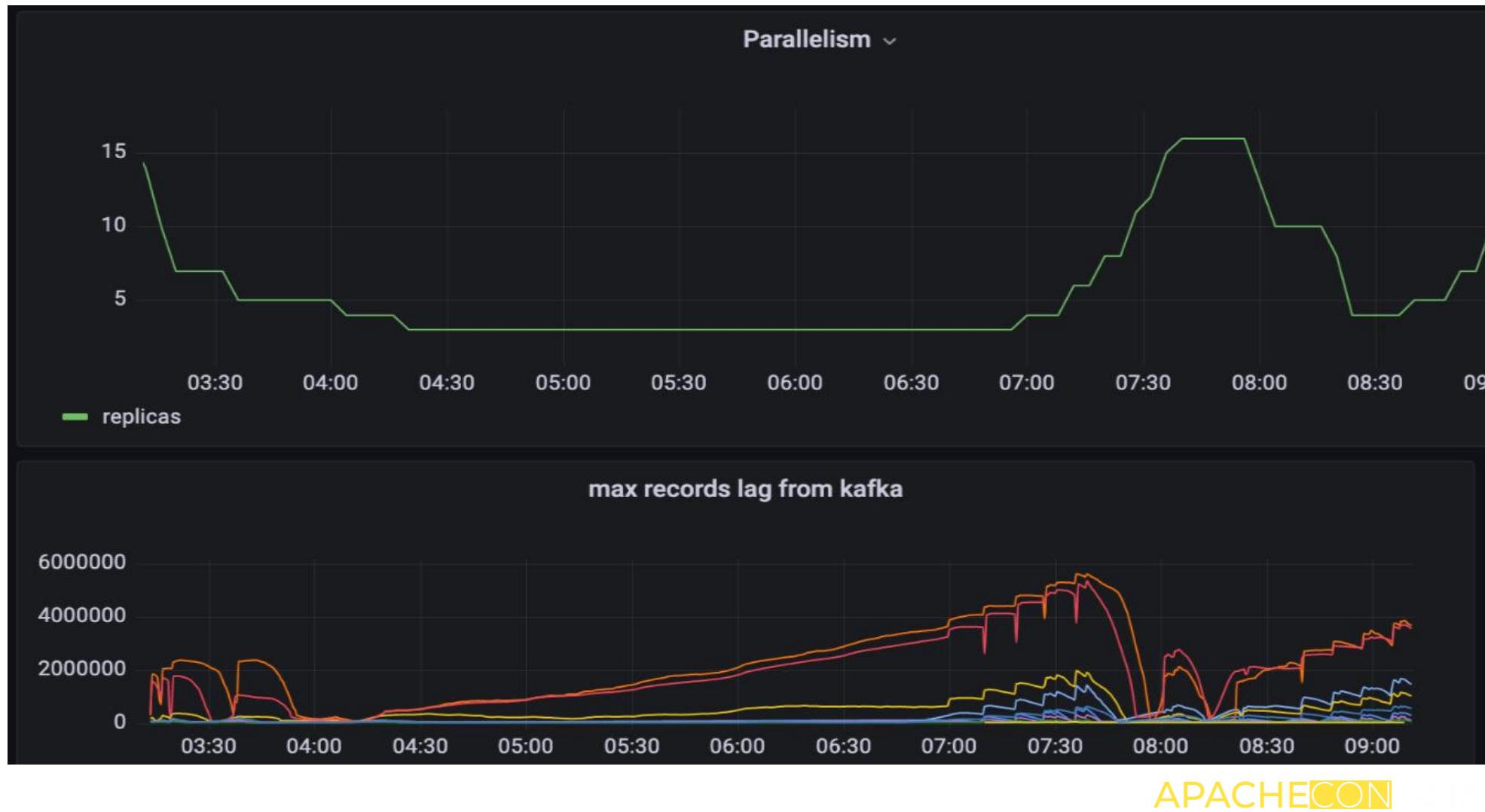
selected mappings: Set Filters

UDF:

Operations:

Pagination: < 1 2 3 4 5 6 ... 89 >

# Simple ETL pipeline – Automatic Scaling



# SQL Task Customization

- Snowflake Support in SQL Task
  - Upsert feature for Snowflake Spark connector
  - Sink selection

Datasource	HIVE/KYUUBI	KYUUBI PROD
SQL Type	Non Query	Operation mode
SQL Parameter	Please enter format key1=value1;key2=value2...	
SQL Statement	<pre> 1 SELECT 2   [REDACTED] 3   [REDACTED] 4   [REDACTED] 5   [REDACTED] 6   [REDACTED] 7   [REDACTED] 8   [REDACTED] meetings 9 FROM mysql 10 WHERE day&gt;= '\${firstDay}' and day&lt;= '\${yesterday}' </pre>	
Selected Columns	[REDACTED]	
Unique Keys	day,hg_id	
target database	SNOWFLAKE	
target table	COLLAB_DB.COLLAB_MEETINGS	
UDF Function	Select	

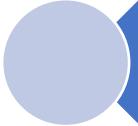
# Resource



website: <https://dolphinscheduler.apache.org>



Github: <https://github.com/apache/dolphinscheduler>



E-mail: [dev@DolphinScheduler.apache.org](mailto:dev@DolphinScheduler.apache.org)



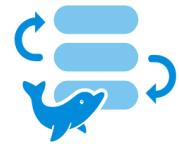
Slack: <https://s.apache.org/dolphinscheduler-slack>



Twitter : [@dolphinschedule](https://twitter.com/@dolphinschedule)



Demo: <http://106.75.43.194:8888/>



# Apache DolphinScheduler

## Smart, Easy and Stable Data Job Orchestration Tools



### Q & A

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