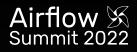
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AIRFLOW SUMMIT





Scalable and Isolated Dag Development Experience at Lyft

&

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Paul Dittamo





Kyte



What is our scale

Users

- Kyte 60 WAU
- Airflow > 350 Active users
- Over 1000 lifetime users (including attrition)

DAGs

- 4000 DAGs on airflow
- 2500 Active DAGs
- Some DAGs have >1000 tasks



Problem Statement

What Problems were we solving?

Lots of requirements for our Airflow Instance

- Python packages
- Environment Variables
- Airflow Configurations

Pipeline development requires production data

- Mocking data is hard
- Staging data is too random
- Statically defined data becomes out of sync with upstream

Wide Variety of Users

- Some technical (SWE)
- Some not (Legal Compliance Teams)
- Need them all able to test DAGs



Implementation

What is Kyte?

Built on existing Lyft platform

- <u>Lyft internal container orchestration</u>
 <u>service</u> (ML Model Training + Batch
 Prediction)
- Kyte container mirrors Airflow production containers

Additional Details

- Sync DAG changes with git
- Wrapper commands to test DAGs
- No scheduler



Key Features



Isolation

Metastore Isolation File System Isolation Performance Isolation



Ease of Use

Utilities Pre-installed Persistent State Jupyter Notebook Support



Read access to production resources

Users write to personal schemas IAM roles + Envoy routing



Key Features Platform/Support



Ease of Support

Remote Development Environment = 100% Visibility + Commoditized Instances



Auditable

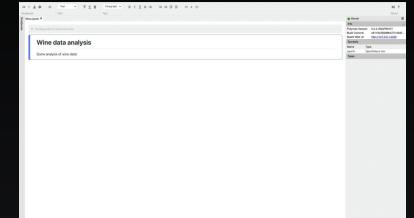
Given the access to production data, it's important to have a trace of the work 6

Historical use case support

Build the new while supporting the old

Inspired by Google Collab and Polynote

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{x}	Setting started Nata science Machine learning Afore Resources Featured examples	 Data science With Colab you can harness the full power of popular Python libraries to analyze and visualize data. The code cell below uses numpy to generate some random data, and uses matplotlib to visualize it. To edit the code, just click the cell and start editing. 		Wir Some i	
	readined examples	<pre>[] Import mampy as np from matplotlh import pyplot as pit y = 300 + np_random.randn(10) x = [x for x in range[lan(r#)] plt.ploty.rym, '-') plt.fill_between(x, ys, 35, where=(ys > 355), facecolor='g', alpha=0.6) plt.title("sample Visualization") plt.title("sample Visualization") plt.tovor)</pre>			
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Demo

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Learnings What did we learn?

Users don't love Github as a syncing mechanism

- Users might be unfamiliar with Github syntax
- Uncomfortable committing
 incomplete code
- Some context switching between environments

Varied preferences for developing locally versus remotely

- Jupyter incomplete as IDE
- Remote environment introduces
 lag
- Setting up IDE can be a cost

Testing failure cases can be difficult

- SLA misses
- Upstream failures
- BranchPythonOperator cases



Next Steps Future Work

Testing Support

- Failure cases
- Integration testing of DAGs

Connect the Notebook

Kernel locally

• Improve local development

Linting/Static Analysis

- Already some work for this open source
- Errors introduced when SQL is copied over and parameterized
- Blessed DAG structure



The Dream Northstar

As Jupyter Notebooks became to Data Analysis, Kyte will become to Airflow dag development

