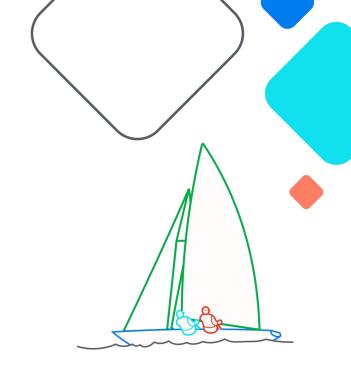
# Democratizing **ML** feature store framework at scale

Rafay Aleem & Victoria Varney



# **XAirflow Summit**

Let's flow together

September 19-21, 2023, Toronto, Canada



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## Today we are talking about....

Democratizing a machine learning feature store framework at Faire

How we have enabled everyone to contribute to a shared resource easily

Building a framework
on top of Airflow by
leveraging it's lowlevel APIS

How Airflow is much more than just a workflow management tool

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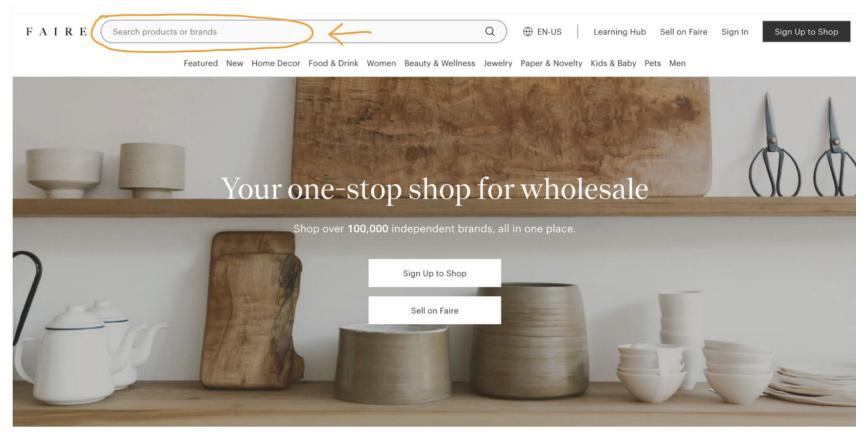
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# **Search experience at Faire**



Storage V

High sell-through

MSRP \$9.83 ⊕

Home / Food & Drink

#### Wholesale Food & drink

High sell-through

FREEZE DRIED JOLLY

MSRP \$9.95 🖯

Upscale Freeze

BALLS (JOLLY RANCHERS)

Shop independent brands from around the world at wholesale prices.

Location ~

Lead time V

High sell-through

2oz Sea Salt Chips (case of

1 in 6 Snacks- Carolina...

20 bags)

MSRP \$49.15 A

Minimum V





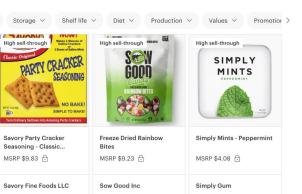
Beverages

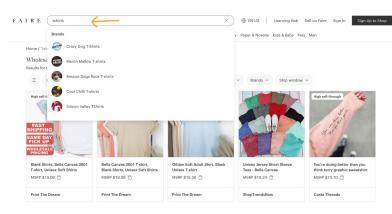


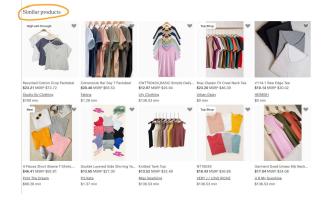




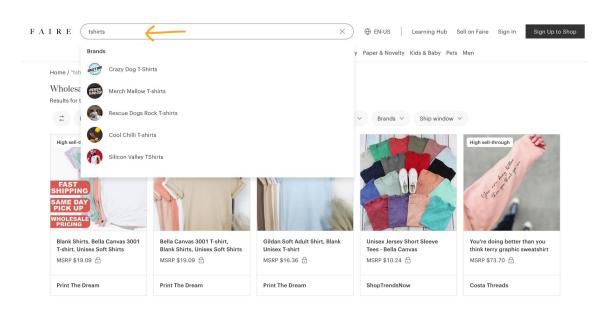


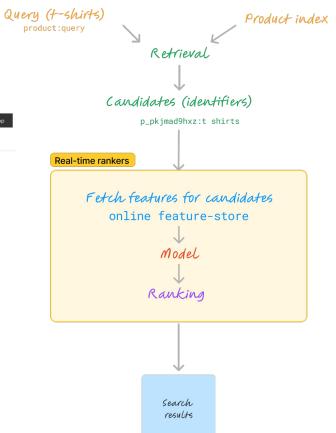




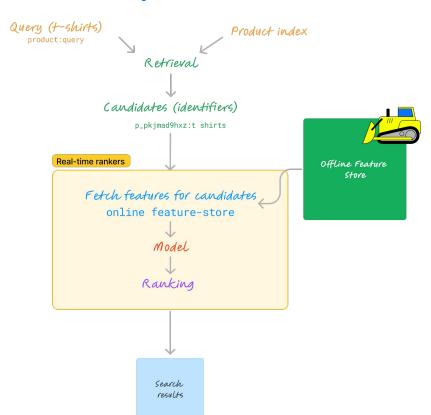


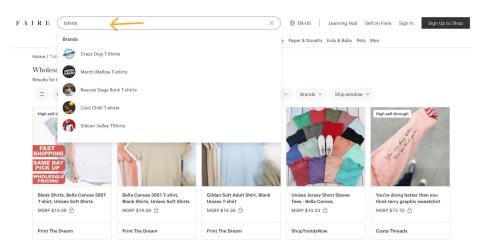
### **Anatomy of a search**





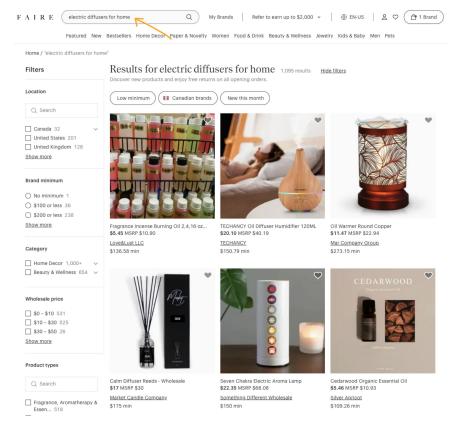
### Anatomy of a search





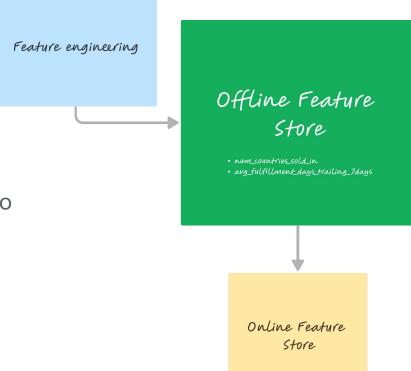
### **Enhancing the search experience**

- Unlock the science behind search relevancy
- Search is particularly important in context of e-Commerce

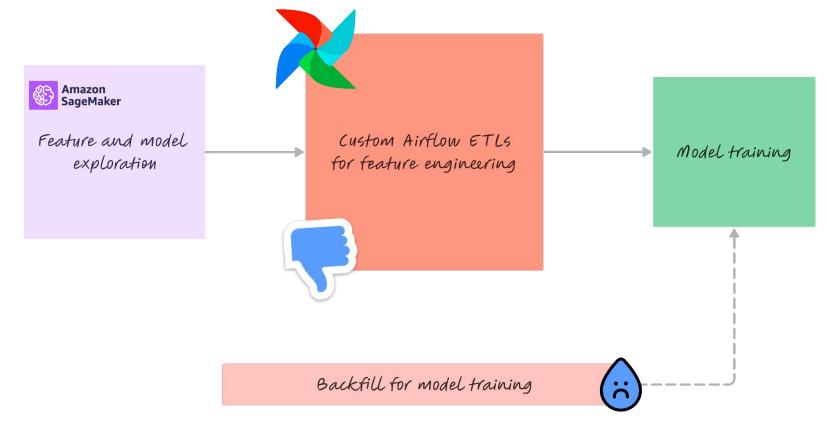


## **Enhancing the search experience**

- Features are float values with some underlying meaning such as
  - num\_countries\_sold\_in
  - avg\_fulfillment\_days\_trailing\_7\_days
- Features are engineered and defined on the offline and eventually propagated to online feature store

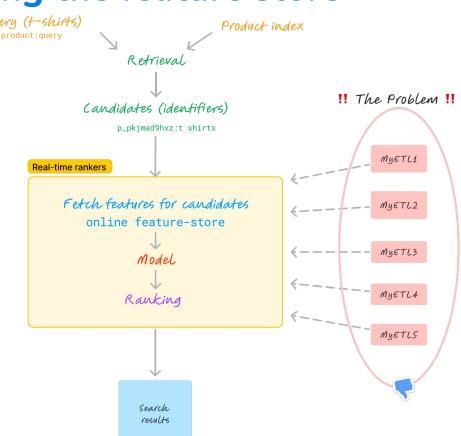


# The need for democratizing the feature store

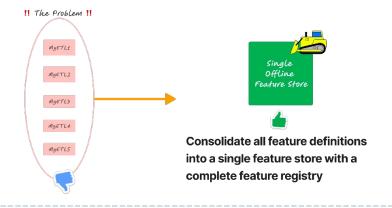


## The need for democratizing the feature store

- No **feature visibility** across teams
- No clear process to separate
   online features from offline
- Non-standardized error-prone backfilling with no clear notion of point-in-time joins
- Overhead costs on ElasticCache and Snowflake



## The need for democratizing the feature store







### **Feature Store Framework**

- All features are defined as SQL with their
   Python configurations
- Each feature can be configured separately with its own metadata

#### Examples

- specify if available online
- feature description
- author
- All configuration is fed into a queryable feature registry

```
b.token as brand_token,
    count(distinct rb.retailer_id) as number_of_contact_books_found
from production.retailer_to_brand_email_domains rb
join production.brands b
    on b.url_domain = rb.email_domain
where b.active
    and rb.created_at < '{{ ds }}'::DATE
group by 1</pre>
```

```
feature groups = [
    FeatureTask(
        sql file="brand contact book.sql",
        features=[
            Feature(
                sql_feature_name="number_of contact_books_found",
                feature description="Number of retailers with this brand in
their address book",
                available online=False,
        entity=BrandEntity(),
        author="rafay",
        is_static=False,
brand_features = FeatureEntity(entity=BrandEntity(), tasks=feature_groups)
```

### **Feature Store Framework**

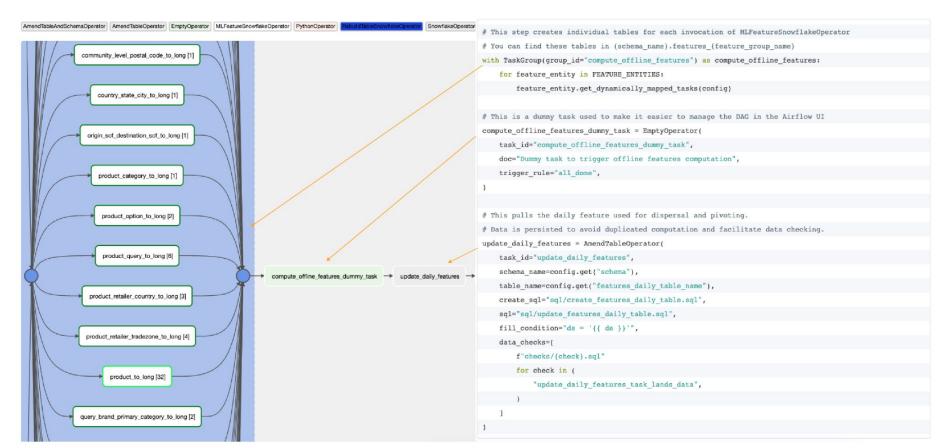
```
class FeatureEntity:
   def __init__(
        self,
        entity: EntityDescription,
       tasks: list[FeatureTask],
        self.entity = entity
       self.tasks = tasks
        self.name = entity.get feature name string()
        for task in tasks:
            assert type(self.entity) == type(task.entity)
            self.check_columns_present_in_sql(task)
        # check dup feature names
        self.check dup feature names()
   def check_dup_feature_names(self) -> None:
       Check whether feature names are unique.
   def get compute operator(
        self, task: FeatureTask, is backfill=False, start date=None, end date=None
    ) -> MLFeatureSnowflakeOperator:
```

#### Low-level APIs

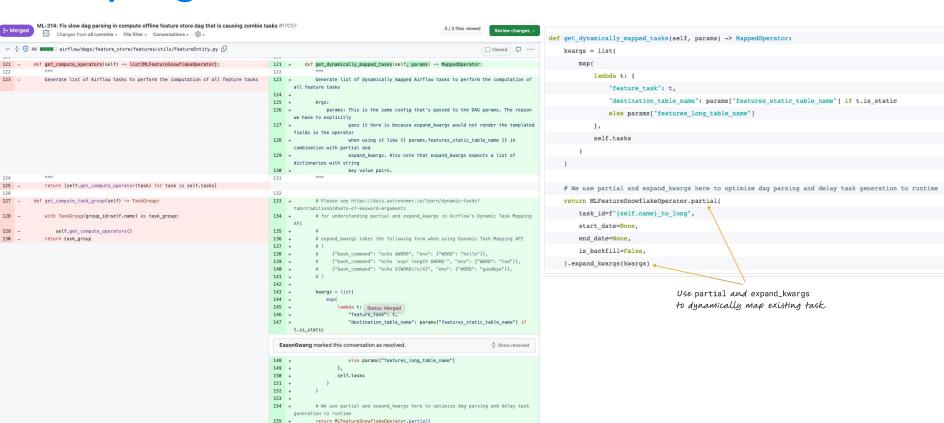
```
Adataclass
class Feature:
    sql feature name: str
    feature description: str = "No description has been specified"
    available online: bool = False
@dataclass
class FeatureTask:
    sql file: str
    features: list[Feature]
    entity: EntityDescription
    author: str = "Unknown author"
    is static: bool = False
    # backfill parameters
   backfill config: FeatureTaskBackfillConfig = None
```

High-level config dataclasses

### **Feature Store Framework**



## **Adopting new Airflow features**



task\_id=f"{self.name}\_to\_long", start\_date=None,

## **Adopting new Airflow features**

- Be **proactive** in adopting new versions of Airflow
- New features have greatly **improved** cluster performance
- Notable features:
  - Dynamic Task Mapping
  - Task Groups (easier visual dag management)
  - **Deferrable operators** for long running tasks such as AWS Batch jobs

### **Extending the framework for feature backfills**

```
FeatureTask(
    sql file="brand contact book.sql",
    features=[
        Feature(
            sql feature name="number of contact books found",
            feature description="Number of retailers with this brand in their address book",
            available online=False,
            backfill config=FeatureTaskBackfillConfig(
                start date=datetime(2023, 8, 1),
                end date=datetime(2023, 8, 30),
                namespace="brand 1 month backfill",
    entity=BrandEntity(),
    author="rafay",
    is static=False,
```

feature\_store\_backfill\_brand\_1\_month\_2023\_8\_1\_2023\_8\_30

- · Uses same SQL files
- Provides extra jinja templated flag {{ is\_backfill }}
- Entire process takes 3 lines of configuration!

```
def get compute operator(
   self, task: FeatureTask, is_backfill=False, start_date=None, end_date=None
) -> MLFeatureSnowflakeOperator:
   Generate Airflow task to perform the computation of a feature task
        task: FeatureTask object
        is backfill: specifies whether the operator should be generated for backfill
        start date: start date of the task, None implies it will match the dag start date
        end_date: end date of the task, None implies it will match the dag end date
   destination_table_name = {
        "{{ params.features static table name }}"
        if task.is static
        else "{{ params.features_long_table_name }}"
   return MLFeatureSnowflakeOperator(
        task id=task.sql name,
        feature_task=task,
       destination_table_name=destination_table_name,
       start date=start date.
        end date=end date,
        is backfill=is backfill,
    ) # type: ignore
```

# Best practices for designing extensible frameworks

 Use mixins to extend operator capabilities

class RebuildTableSnowflakeOperator(
SnapshotMixin, TableOperatorMixin, ETLSnowflakeOperator

capabilities using mixins

 Allows adding new features to low-level
 APIs without breaking

```
class SnapshotConfig:
    Frequency = Enum('Frequency', ['NEVER', 'DAILY', 'EVERY_RUN'])

store_daily_embeddings = RebuildTableSnowflakeOperator(
    task_id="store_daily_embeddings",
    sql=daily_online_embedding_sql(),
    schema_name=config.get("schema"),
    table_name="daily_embeddings_v2",
    snapshot_frequency=SnapshotConfig.Frequency.EVERY_RUN,
)
```

# Airflow beyond workflow management

Airflow goes far and beyond a workflow management tool

It's thoughtful design makes it very extensible and powerful

It is very well suited for running mission critical workflows with tight SLA requirements

It continues to be a very stable part of Faire's infrastructure and continue to scale

### **Key Takeaways**

Proactively consider onboarding to Latest Airflow versions

Don't think of
Airflow as just a
workflow
management tool

Consider building
shared
frameworks
instead of shared
ETLs

Thoughtful use of Airflow APIS and features goes a long way

### **Credits**

- Wayne Zhang for his guidance on the offline feature store framework
- Analytics Engineering team for their feedback on table snapshot tooling
- Core Data Infra team for their constant support with Airflow and Snowflake
- Machine Learning Platform team for dealing with on-call issues and providing stakeholder support
- My wife's constant support

## Wrap Up & Questions

Careers @ Faire: <u>faire.com/careers/</u>

- Where to find Rafay
  - LinkedIn: <u>linkedin.com/in/mrafayaleem/</u>
  - Airflow Slack: <u>@Rafay Aleem</u>
  - Twitter: <u>@mrafayaleem</u>
  - Email: <a href="mailto:contact@mrafayaleem.com">contact@mrafayaleem.com</a>
  - Newsletter: <u>mrafayaleem.com/#/portal/</u>



