# coalesce



Scalable data pipelines that empower financial analysts

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#### Meet today's presenter from this company





Data Engineer

Tech Lead

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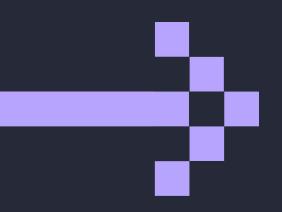


#### How to build cross functional data teams

# Business Value = Engineering + Subject Matter Expertise







# Background



#### dv01 is the Data Hub for Institutional Investors in Structured Finance



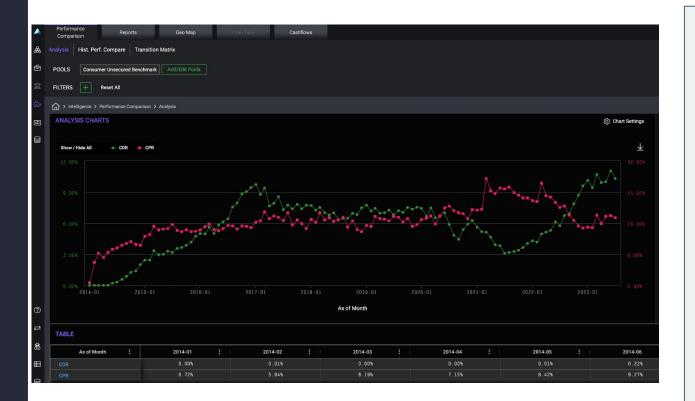
- A retail investor:
  - □ **Security Types**: Stocks, ETFs, Mutual Funds, bonds
  - □ **Performs:** Research and analysis
  - □ **Why:** To determine which to buy and sell

- A structured finance investor has the same motive as retail investor but the securities comprises pools of loans:
  - Security Types: Asset-Backed Securities ("ABS") & Mortgage-Backed Securities ("MBS").





## The dv01 Web App

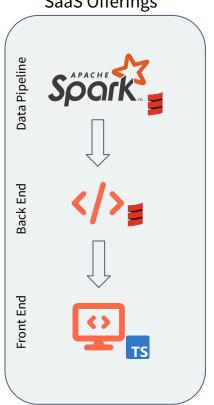


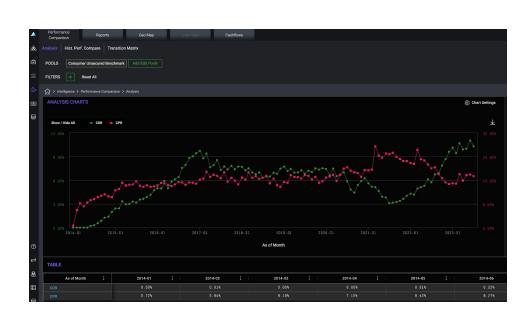
- dv01
   standardizes
   loan data and
   offers users
   embedded
   analytics for
   analysis
- Data can also be accessed via SFTP files and BigQuery



# Our data pipeline is designed for a highly engineered tech stack

#### SaaS Offerings





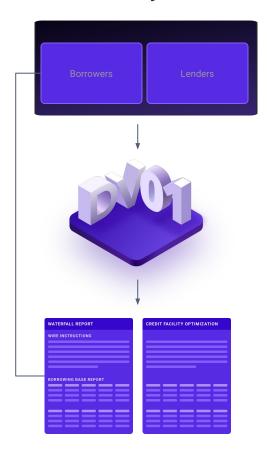


# A new business line sprouts





#### A credit facility is credit line for investors offered by a bank

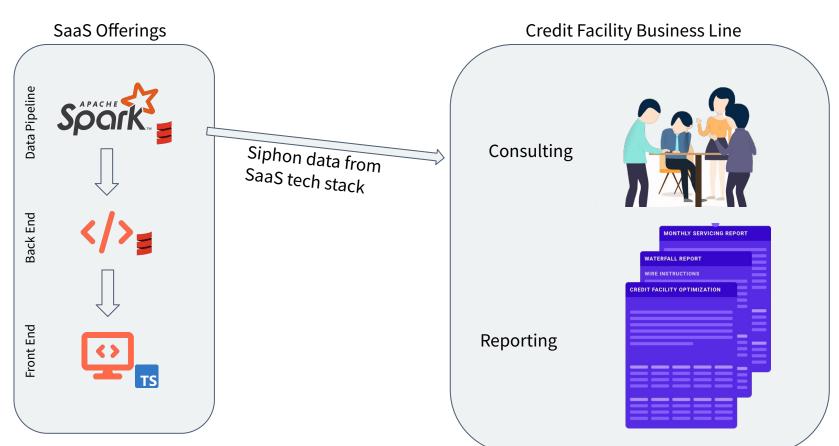


A borrower negotiates a revolving credit line from a lender. The borrower uses the credit line to purchase loans. The purchased loans become collateral used to make additional borrowing from the credit line.

- Agreements include stringent requirements on asset performance and financial covenants
  - Ex: The collateral pledged to a facility cannot have more than 20% of loans issued in a given state, delinquencies cannot exceed 10%
- Borrowers must produce reports to lenders that prove that facilities meet requirements



# dv01's data workflow is designed for a front end, not bespoke reporting



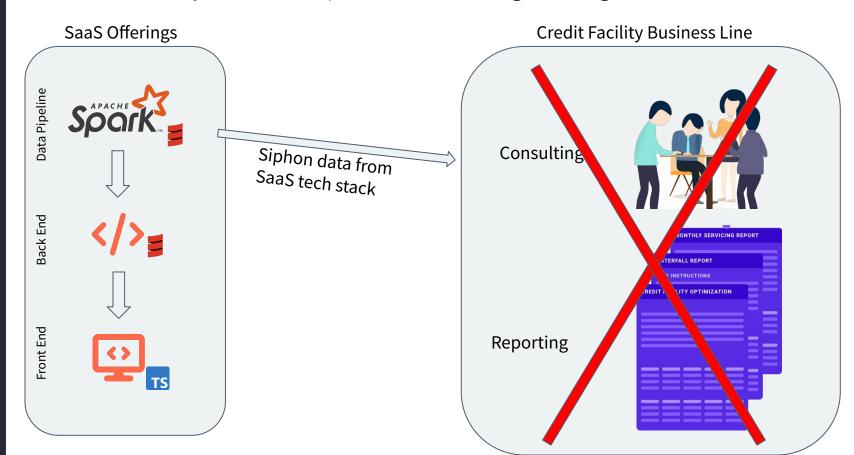


# A new business line grows



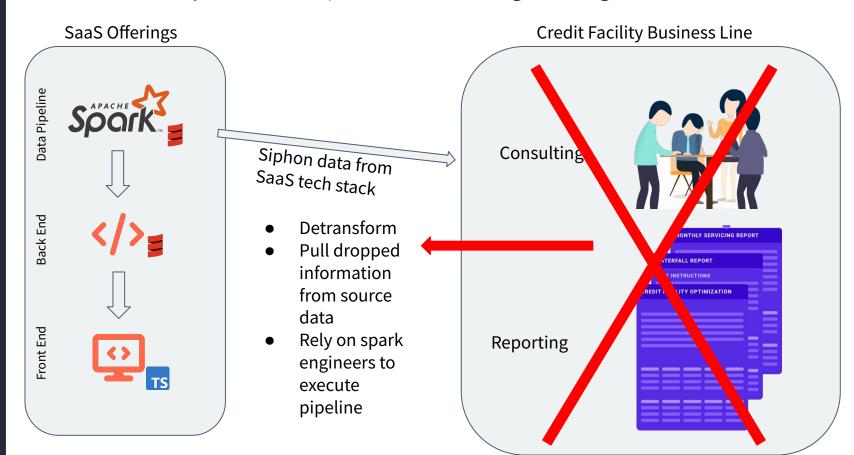


## Financial Analysts are co-opted into data engineering





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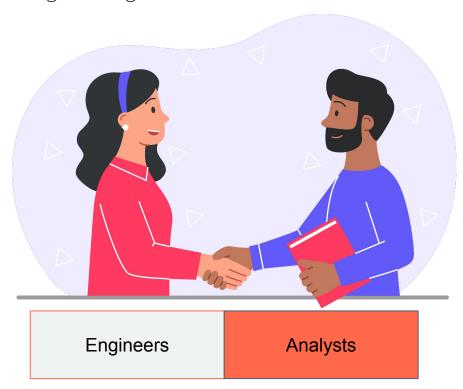
**Problem:** Spark data pipeline was built for SaaS offering which had conflicting requirements with our new business line

- SaaS Offering
  - Standardized: Allows apples-to-apples comparison
  - Regular schedule: Set cadence of data updates
  - Owner: Engineers

- Reporting Services
  - Tailored: Analysis based on idiosyncrasies of each facility
  - □ **Episodic schedule:** Each facility has unique requirements
  - □ **Owner:** Analysts

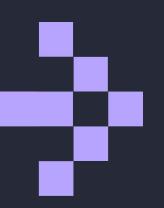


**Solution:** Build a team heavy on subject matter expertise with enough engineering to be self sufficient



- SQL
  - Language usable by engineers and financial analysts
- dbt
  - flexible orchestration
  - Transparent data pipelines

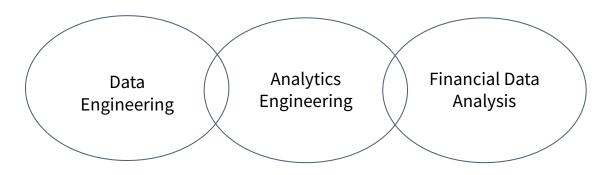




# Credit Facilities Team Composition



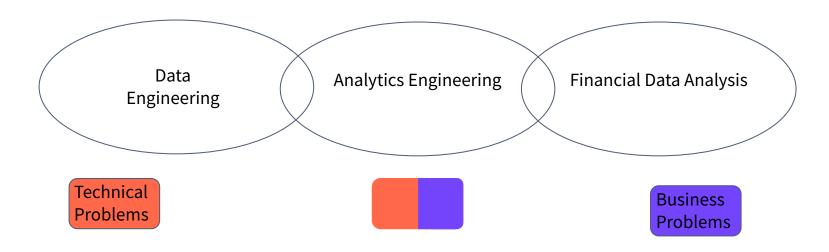
#### **Job Functions**



#### **Technologies**







- Distinct roles with some overlap
- Overlap allows a lean team to handle volatile workloads





- Focused fully on business problems
- Significant client contact
- Trained to understand dbt codebase and ad hoc SQL queries
  - □ Do not make major contributions to dbt codebase
- Freed from juggling different roles





- Work end to end within dbt pipeline
- Implement business logic and tests in SQL & dbt
- Recruited from data analysts with ambition to grow technically
- Possess significant SME and have some client interaction
- Example technical responsibilities
  - □ Code review and approval processes (PRs)
  - □ Manage CI/CD
  - Manage orchestration



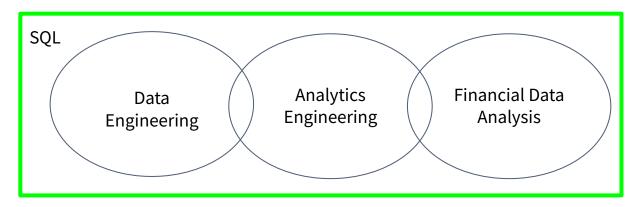


- Some overlap with analytics engineers on day to day
  - Focus engineering heavy problems, allowing analytics engineers to focus on the confluence of business and engineering
- More collaboration with engineers from other parts of the company than analytics engineers
- Mentored on SME by financial analysts
  - □ Less SME compared to data analysts and engineers but broader technical skills



# Why SQL + BigQuery?

#### SQL breaks down silos



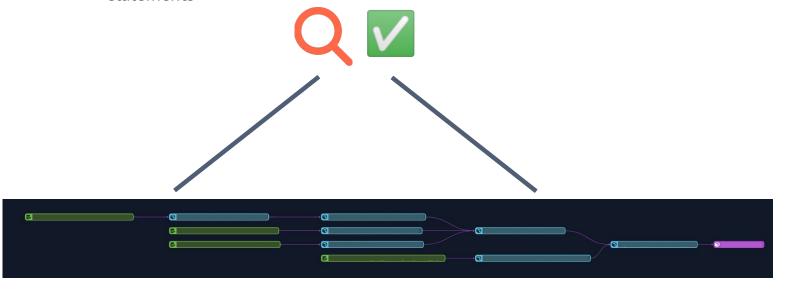
BigQuery consolidates source and final data in a single location



#### Why dbt?

- dbt makes SQL organized, modular and testable
- dbt models make intermediate states transparent
  - It is challenging to inspect intermediate states of lazy evaluated spark pipelines, often requiring a debugger and expensive collect statements









Laying the foundation for a cross functional data team



## SQL Training

#### Goals

- Teach all analysts to interrogate data with SQL
- Provide foundation for analysts who want to go deeper

Migrated to
BigQuery
Workflows
Tailored
Curriculum

Start Course
Curriculum



# SQL Curriculum Guiding Principles







**Practical examples** 



Cookie-cutter exercises



## Growing Analysts to Analytics Engineers

#### Why

Niche domain knowledge

Technically sophisticated analysts

#### Who

Analysts from SQL training with:

- 1. High technical aptitude
- 2. Interest

#### How

**Mentoring Program** 

Analysts <> Experienced Engineers



#### Keys to successful mentoring program

# **Incentive Alignment**

Recognize mentors in performance reviews

# **Core Competencies**

Performant SQL

Code Review & PR
Practices

CICD!!

#### Personal Learning Plans

Mentors help design

Mentees engage in self study

#### Pair Programming

I do

I watch you do

You do



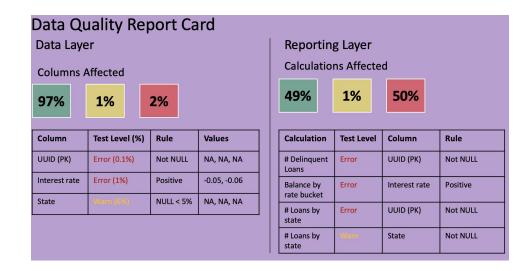
#### Instituting a Culture of Test Driven Development

- Testing starts with analysts
  - □ Best equipped to define business logic assertions and nuanced client requirements
  - □ Analysts suggest tests for every client requirement
- Engineers implement tests
  - □ Implement tests on sources and models
  - Set of standard column level tests & relational tests on dbt DAG



## Impact of Testing Culture

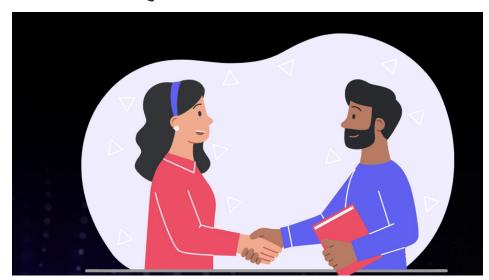
**Reactive** Proactive





#### Conclusion

- Business Value = Engineering + Subject Matter Expertise
- SQL unifies domain knowledge & engineering
- dbt makes SQL modular and testable



Slides available at





# Thank you

