# How to reduce your dbt data pipeline costs with DuckDB



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#### Introduction

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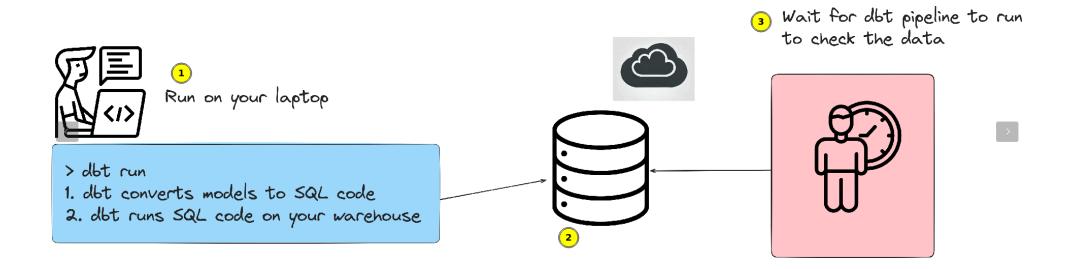




### [Why] DuckDB + Ephemeral VMs = dirt cheap data processing.

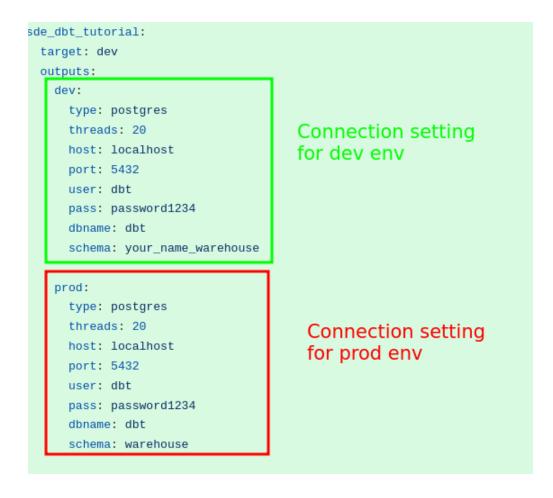


#### Standard dbt + warehouse workflow





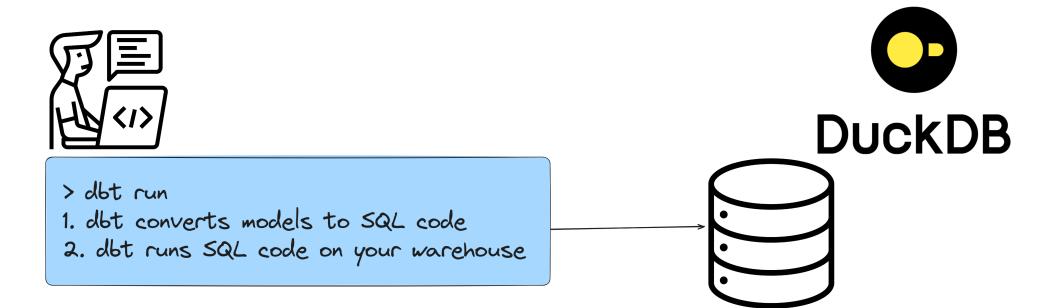
• The connection settings are defined in the profiles.yml file.





### dbt + duckdb local workflow

Run on your laptop





### Serverless data pipeline workflows

 Serverless dbt + duckdb workflow: Start VM -> Pull data into VM -> Process data -> Dump data into destination -> Spin down VM.



Note: Pulling data can be done effectively with DuckDB extensions



### Ephemeral servers are inexpensive

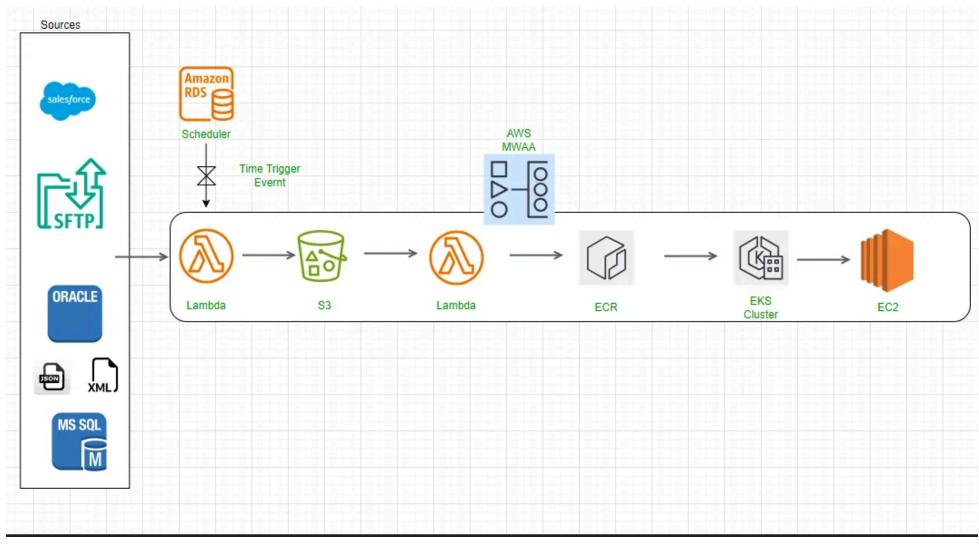
- Most cloud providers allow you to rent virtual machines charged by the hour (e.g., EC2).
- If you are not under tight time constraints, you can bid spot VMS at a much lower rate than on-demand VMS.
- Pulling data into VM (e.g. EC2) is cheap.
- Fast data transfer speed (especially with cpp optimized duckdb extensions).



### Real project cost computation

- Real data infra (ref). Details:
  - 1. **Number of jobs**: 800 independent data pipelines with about 80 source systems
  - 2. Number of files: ~400
  - 3. Average size per file: ~1GB
- Objective: Reducing overall costs to between \$3,000 and \$5,000 per month, processing about 400GB per day across 800 jobs.





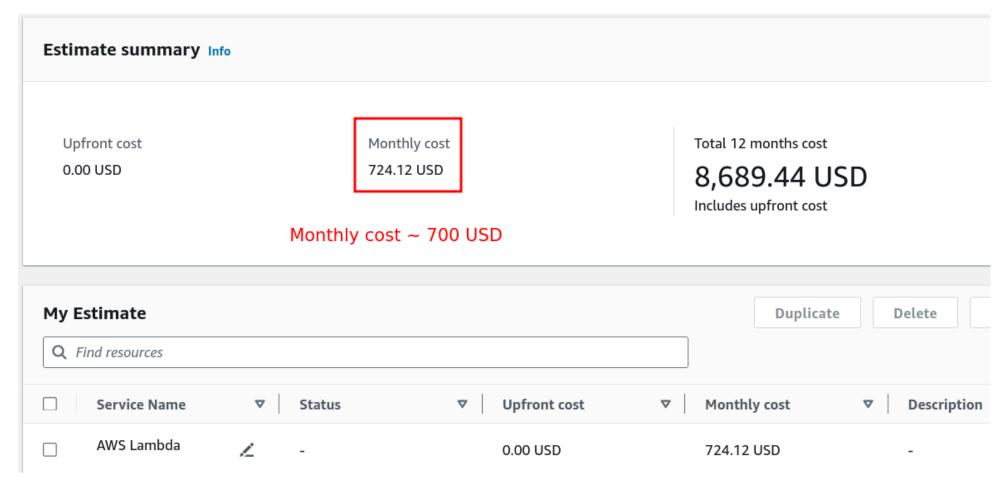
Architecture



### Cost projection with serverless duckdb + dbt

- Execution time (sample): A non-optimized, expensive pipeline (fact-fact join, joining all dimensions) takes about 2 minutes to run.
  - 1. Input: ~2GB
  - 2. Github codespace machine: RAM: 8GB & Cores: 2
  - 3. Output: ~13GB
- Serverless for 800 jobs ~ 700 USD.
- Data transfer costs & IP costs are negligible at this scale.





https://calculator.aws



### Save time and money by keeping the feedback loop short

- With standard dist. Data proc systems, you will need to wait a while (set up cluster, start job, etc.) before you see the results.
- With dbt + duckdb, you can run locally without a complex setup and see results instantly (check out the buenavista package for viewing results when a dbt pipeline is running in duckdb).



#### Recap

- Cost saving by avoiding pay-by-data-scanned cost model.
- **Time saving** by reducing feedback loop and avoiding paying for development.



[How] Migrate one (or a few related pipelines) at a time.



### Your first migration will involve some work

- Setting up infra to run, EC2/AWS Lambda/ECS, k8s, etc instead of connecting to a db/engine.
- Code changes: Changing db-specific functions, e.g., merge into, date functions, etc.
- Input data: You have to read the data from source systems or from warehouse where the input data exists.



- Error handling & Debugging: If you are processing in memory (without persisting intermediate datasets) and your data processing fails, you will have to re-run the entire pipeline.
- Permissions are set at the service level, not the USER/ROLE level, as in most data processing systems.
- Logging system metrics: most cloud VMs have this setup,
   e.g., AWS Cloudwatch. In addition to metrics logging you will not be able to see query history.



### With a template to migrate, the rest of the migration will be more straightforward

- With the infra in place, the migration will be simple.
- Migrate at off-peak times. For example, if you are in e-commerce, don't migrate during Thanksgiving, or if you are in finance, don't migrate at the end of fiscal year reporting.



- Data pipeline migrations are tricky!
  - 1. Code: Ensure the code has unit/integration tests (not just DQ checks)
  - 2. **Data**: Ensure data has sufficient DQ checks and validate data between old and new systems for a defined period before switching over.

### Watch out for pipelines that fully reprocess huge tables.

- If you have pipelines that involve aggregating historical data (e.g., for anomaly detection, handling late-arriving events), you need to handle them. Use one of the methods below:
- Aggregate past n periods of data instead of reaggregating the entire data set. For example, if your pipeline processes sales data that comes in every day, instead of reprocessing the past n years' worth of data, consider reprocessing the past 3/6 months of data (depending on later arriving data for your business use



• Store aggregated data in a separate location. For example, if you count rows in a dataset every run, store the counts in a separate table so you don't have to recompute them for historical data each time.



#### Multi environment setup

- There may be cases where you may want to run some pipelines with duckdb and the rest with your existing warehouse.
- You can use dbt's profile.yml and dbt cli to indicate which pipeline should run where.
- Note that this introduces significant complexities with debugging & maintanance & establishing SOT!
- But you can alleviate a lot of issues if you are using a catalog (e.g. iceberg catalog) that most OLAP dbs support.



```
(env) → how-to-slash-dbt-cost-w-duckdb git:(main)
                                                        ✗ dbt run --target dev
config:
                                                       (env) → how-to-slash-dbt-cost-w-duckdb git:(main)
 send_anonymous_usage_stats: false
sde_dbt_tutorial:
                                                   918 X dbt run --target dev-cloud-warehouse
 target: dev
 outputs:
   dev:
     type: duckdb
     path: ./dbt.duckdb
   dev-cloud-warehouse:
     type: postgres
     threads: 20
     host: ecs.xxxx.com
     port: 5432
     user: dbt
     pass: password1234
     dbname: dbt
     schema: warehouse
```



## [Caution] Watch out for integration points!



### Data permissions are defined at the service level.

- Most data processing systems have comprehensive data access controls crucial for data governance.
- With dbt + duckdb, we must handle data access at a service (E.g., AWS Lambda can access a specific S3 bucket, etc.) level.
- The inability to handle data permissions at a row level (like what Snowflake can offer) can sometimes be a deal breaker (PII, Sensitive info, etc).



### Some popular tools don't officially support DuckDB yet

- 1. elementary (GH issue)
- 2. greatexpectations



### You need to do some work to dump data into Vendor warehouses

- With dbt, you usually create the output dataset as well.
- When using duckdb for processing, you must dump the data into a destination system (cloud store or another database).
- You will need a system to create the output tables/dump to the cloud store via extensions or code.



#### **Excited About**

- 1. Motherduck makes serverless infra for proc and analytics simple
- 2. Ibis dataframe with DuckDB backend
- 3. Tight Python integrations with Client API
- 4. dbt duckdb
- 5. Buena vista library for accessing duckdb tables when dbt is running



### Questions

