6/10/25, 3:04 PM Sigma



QUERY_HISTORY

Query Id	Query Text V	Database Id 🗸	Database Name 🗸	Schema Id 🗸	Schema Name 💙	Query Type 🗸	Session
64ebe479-0501-40aa-0004-ba03064a61ba	select * from null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SELECT	13304219728
c6a06f47-0501-40aa-0004-ba030649b50a	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219728
3e43ee1c-0501-405d-0004-ba03064a0bce	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219728
bc8468df-0501-405d-0004-ba03064a0bfe	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219728
3dde65b0-0400-cf74-0004-ba030550af66	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219660
a03bb09a-0400-d0ee-0004-ba0305519c22	select * from null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SELECT	13304219661
3e60e75d-0400-cfef-0004-ba030550691e	commit	null	null	null	ACME_SCHEMA_BASE	COMMIT	13304219660
3a8a94ec-0601-1830-0004-ba03060a4a0a	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219707
f50481fb-0401-65da-0004-ba03069f9326	select * from null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SELECT	13304219748
217ef5af-0501-3ebf-0004-ba030647a606	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219728
5a6a9fdf-0501-3bf4-0004-ba03063fd362	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219726
4fd44992-0501-5d60-0004-ba03068dc8f6	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219743
57f75a83-0400-bb98-0004-ba0305293012	select * from null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SELECT	13304219647
9ef819a2-0400-936e-0004-ba0304dd01b6	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219625
6b413d1f-0600-f184-0004-ba030585956e	select * from null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SELECT	13304219681
61e9f5c8-0600-f1a8-0004-ba030585b382	show objects like X in schema null.ACME_SCHEMA_BASE	null	null	null	ACME_SCHEMA_BASE	SHOW	13304219681
2462fhe2-0500-6e70-0004-he02040e6h26	select * from pull ACME SCHEMA RASE	null	null	null	ACME SCHEMA RASE	SELECT.	13304219604

DATABASE_STORAGE_USAGE_HISTORY

null null null null null null	519966350880 351497675848 370134535672 1002609581 85480697168 584693082169	3440940 31503542310 30088439223 3440940 3440940 47521918605
null null null	370134535672 1002609581 85480697168	30088439223 3440940 3440940
null null	1002609581 85480697168	3440940 3440940
null	85480697168	3440940
null		
	584693082169	47521918605
null		7/321910003
Hull	352986832982	31574481126
null	370115116965	30102406378
null	1002580627	3438044
null	85480668215	3438044
null	519966321927	3438044
null	584779346869	47730846171
null	85480637382	3434961
null	1002549794	3434961
null	519966291094	3434961
null	374085252432	31621434598
llua	61 <u>4</u> 723QR3R02	A799A35527A
	null null null null null null null	null 1002580627 null 85480668215 null 519966321927 null 584779346869 null 85480637382 null 1002549794 null 519966291094 null 374085252432

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WAREHOUSE_EVENTS_HISTORY

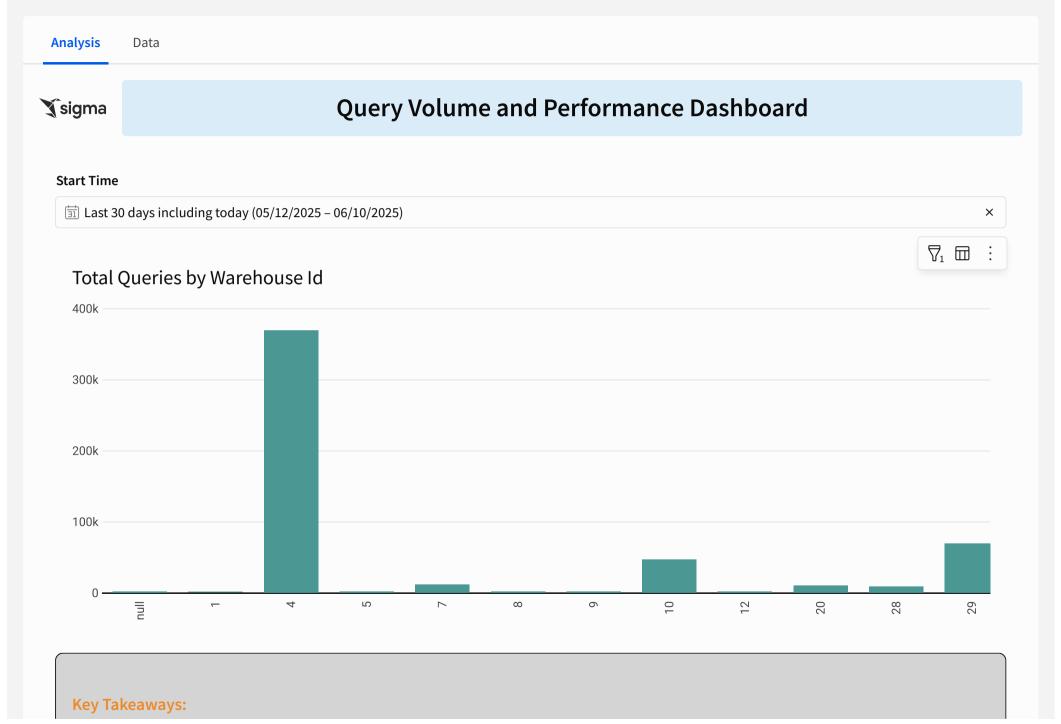
Timestamp ~	Warehouse Id 🗸	Warehouse Name V	Cluster Number 🗸	Event Name ∨	Event Reason V	Event State ∨	User Name ∨	Role Name ∨	
2024-09-07 08:25:32	15	INVENTORY_SCHEDULING_WH	0	RESUME_WAREHOUSE	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:03:04	4	SALES_PROD_WH	0	RESUME_CLUSTER	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:28:35	4	SALES_PROD_WH	0	SUSPEND_CLUSTER	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-07 08:55:19	10	SMOPS_WH	0	RESUME_CLUSTER	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:21:42	4	SALES_PROD_WH	0	RESUME_CLUSTER	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:54:16	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-07 08:31:07	4	SALES_PROD_WH	0	RESUME_WAREHOUSE	WAREHOUSE_AUTORESUME	STARTED	null	null	
2024-09-07 08:36:36	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-07 08:46:26	4	SALES_PROD_WH	0	RESUME_WAREHOUSE	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:08:26	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-07 08:08:25	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	STARTED	null	null	
2024-09-07 08:31:08	4	SALES_PROD_WH	0	RESUME_CLUSTER	WAREHOUSE_AUTORESUME	COMPLETED	null	null	
2024-09-07 08:55:18	10	SMOPS_WH	0	RESUME_WAREHOUSE	WAREHOUSE_AUTORESUME	STARTED	null	null	
2024-09-08 07:08:16	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-08 07:08:16	4	SALES_PROD_WH	0	SUSPEND_CLUSTER	WAREHOUSE_AUTOSUSPEND	COMPLETED	null	null	
2024-09-08 18:37:12	4	SALES_PROD_WH	0	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND	STARTED	null	null	
2024-00-08 22-08-26	4	CVI EC DDUD MH	n	CLICDENIN WADEHOLICE	WADEHOLICE ALITOCLICDEND	COMPLETED	null	null	

WAREHOUSE_METERING_HISTORY

Credits Used C	Credits Used Compute	Credits Used V	e Id V Warehouse Name V	nd Time V Warehouse Id V	End Time V	Start Time ∨
	7.308130961	7.322897778	10 SMOPS_WH	11 11:00:00 10	2025-01-11 11:00:00	2025-01-11 11:00:00
	5.3675534	5.376124345	4 SALES_PROD_WH	12 06:00:00 4	2025-01-12 06:00:00	2025-01-12 05:00:00
	4.285003132	4.291854937	4 SALES_PROD_WH	12 05:00:00 4	2025-01-12 05:00:00	2025-01-12 04:00:00
	4.205722207	4.21260231	8 DATA_ENGINEERING	11 11:00:00	2025-01-11 11:00:00	2025-01-11 11:00:00
	6.871896731	6.882956635	10 SMOPS_WH	11 12:00:00 10	2025-01-11 12:00:00	2025-01-11 11:00:00
	5.807150413	5.824417617	4 SALES_PROD_WH	11 12:00:00 4	2025-01-11 12:00:00	2025-01-11 11:00:00
	7.759536124	7.772005077	4 SALES_PROD_WH	12 07:00:00 4	2025-01-12 07:00:00	2025-01-12 06:00:00
	6.193517946	6.204036994	7 WEBSITE_WH	11 11:00:00 7	2025-01-11 11:00:00	2025-01-11 11:00:00
	4.717844873	4.72537878	4 SALES_PROD_WH	11 18:00:00 4	2025-01-11 18:00:00	2025-01-11 17:00:00
	4.980882102	4.988816742	4 SALES_PROD_WH	11 15:00:00 4	2025-01-11 15:00:00	2025-01-11 14:00:00
	5.758584458	5.768946695	4 SALES_PROD_WH	12 08:00:00 4	2025-01-12 08:00:00	2025-01-12 07:00:00
	7.024720014	7.035810844	4 SALES_PROD_WH	11 22:00:00 4	2025-01-11 22:00:00	2025-01-11 21:00:00
	6.623136803	6.633668003	15 INVENTORY_SCHEDULING_WH	11 11:00:00 15	2025-01-11 11:00:00	2025-01-11 09:00:00
	5.011119743	5.019123823	4 SALES_PROD_WH	12 04:00:00 4	2025-01-12 04:00:00	2025-01-12 03:00:00
	7.827281284	7.842316254	4 SALES_PROD_WH	11 13:00:00 4	2025-01-11 13:00:00	2025-01-11 12:00:00
	5.934296529	5.943806031	4 SALES_PROD_WH	11 19:00:00 4	2025-01-11 19:00:00	2025-01-11 18:00:00
	A 57122262A	A 578383072	4 CVIEC DOUD MH	12 ∩∩-∩∩-∩∩ 4	2025-01-12 00-00-00	2025-01-11 22-00-00

 Data
 Query History Analysis
 Storage Insights
 Events Insights
 Credits Insights
 Warehouse Efficiency Summary

6/10/25, 3:05 PM



6/10/25, 3:05 PM Sigma

Top Performer: The tallest bar is the busiest warehouse. This is where most of our compute demand lives.

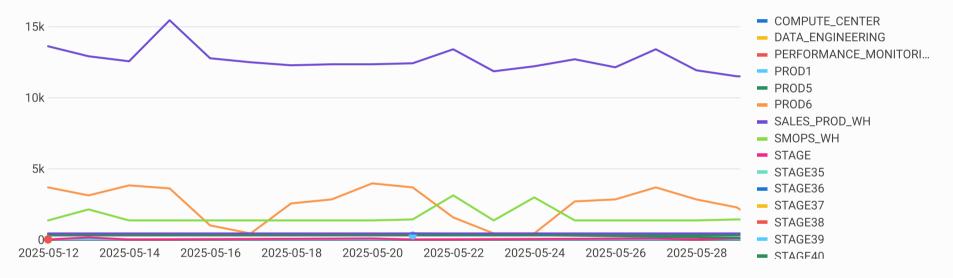
Cold spots: Any warehouses with near-zero bars indicate under-utilized capacity, can be resized.

Start Time

a Last 30 days including today (05/12/2025 – 06/10/2025)

×

Query Volume Trend

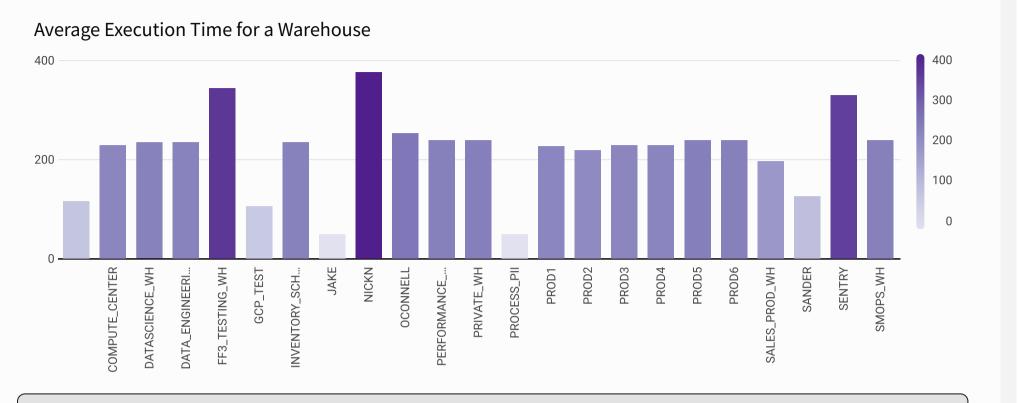


Key Takeaways:

Spikes/dips: Sharp peaks could correspond to a release or batch job; sudden drops might indicate an outage or changed user behavior.

Predictability: A regular weekly pattern (e.g., low on weekends, high on Mondays) tells you when to schedule maintenance or autoscaling policies.

6/10/25, 3:05 PM Sigma



Key Takeaways:

Performance bottlenecks: Warehouses with high average times are slower—maybe under-powered, mis-configured, or querying very large datasets without proper clustering.

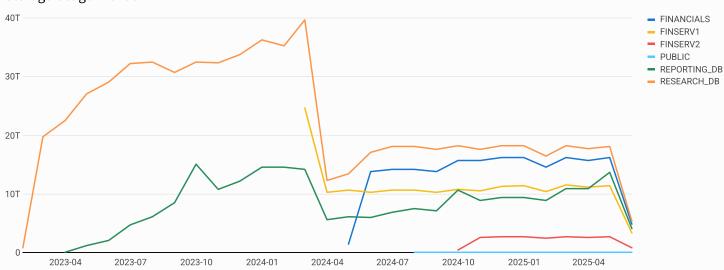
Correlation with volume: If a warehouse has both high volume and high latency, it's a prime candidate for optimization.

Data v Query History Analysis v Storage Insights v Events Insights v Credits Insights v Warehouse Efficiency Summary v

Sigma

Database Storage Growth and Insights

Storage Usage Trends



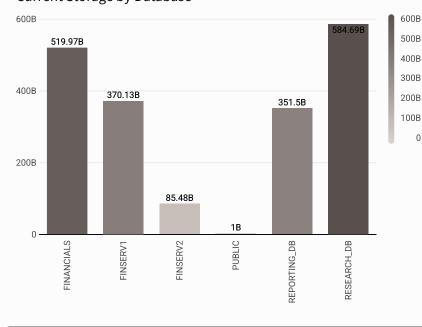
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Key Takeaways:

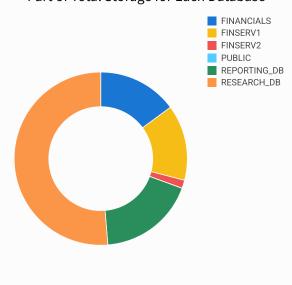
Rapid Growth: Databases with steep slopes may need archiving or partitioning.

Stability vs. Spikes: Sudden jumps could indicate massive data loads or retention policy issues.

Current Storage by Database



Part of Total Storage for Each Database



Key Takeaways:

Cost Hotspots: The top databases drive the majority of storage costs—candidates for compression or data cleanup.

Resource Allocation: Helps prioritize which data stores to optimize first.

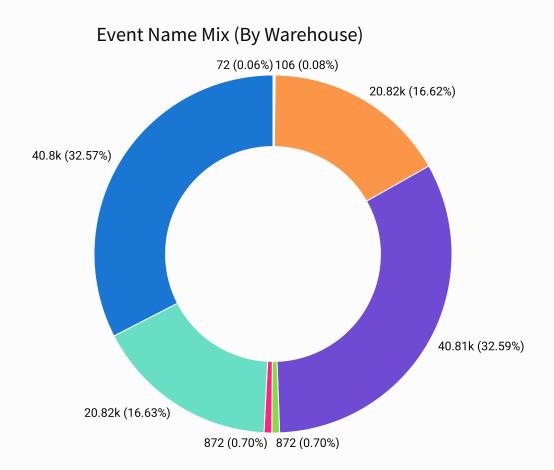
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Analysis

Data



Warehouse Event Patterns Dashboard



Key Takeaways:

We can see that 32.5% events are SUSPEND_WAREHOUSE and 32.5% are RESUME_WAREHOUSE, which indicate frequent autosuspend / autoresume cycles.

ALTER_WAREHOUSE
CREATE_WAREHOUSE

DROP_WAREHOUSE
RESIZE_CLUSTER
RESIZE_WAREHOUSE
RESUME_CLUSTER

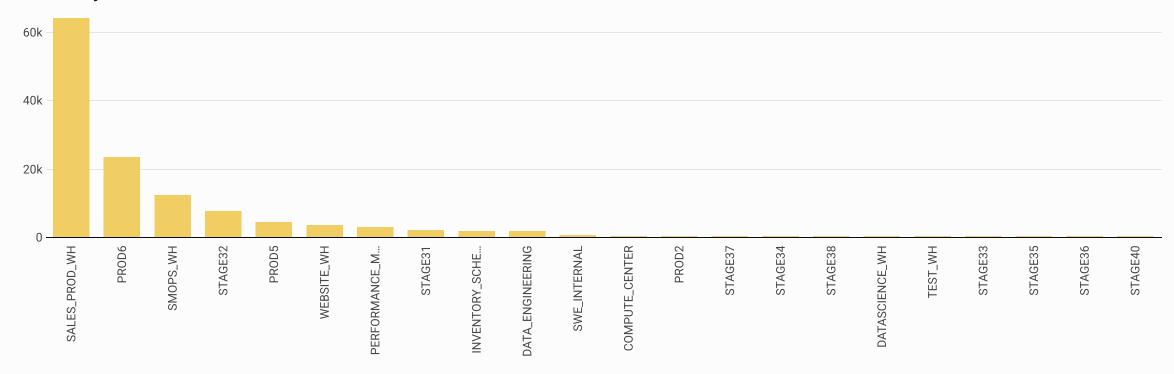
RESUME_WAREHOUSE
SPINDOWN_CLUSTER
SPINUP_CLUSTER
SUSPEND_CLUSTER
SUSPEND_WAREHOUSE

6/10/25, 3:16 PM Sigma

Event Name

(11) RESUME_WAREHOUSE, SUSPEND_WAREHOUSE, SUSPEND_CLUSTER, RESUME_CLUSTER, SPINDOWN_CLUSTER, SPINUP_CLUSTER, RESIZE_WAREHOUSE, ALTER_WAREHOUSE, RESIZE_C... × •

Events by Warehouse



Key Takeaways:

Warehouse SALES_PROD_WH has 64,083 events. Its high event rate may signal oversizing or misconfigured autoscaling. We can investigate its usage patterns and right-size it.

6/10/25, 3:16 PM Sigma

Recent Events

Timestamp ≒↓	Warehouse Name 🗸	Event Name 💙	Event Reason 🗸
2025-06-10 06:19:50	SALES_PROD_WH	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND
2025-06-10 06:19:50	SALES_PROD_WH	SUSPEND_CLUSTER	WAREHOUSE_AUTOSUSPEND
2025-06-10 06:19:50	SALES_PROD_WH	SUSPEND_WAREHOUSE	WAREHOUSE_AUTOSUSPEND
2025-06-10 06:11:41	SALES_PROD_WH	RESUME_CLUSTER	WAREHOUSE_AUTORESUME
2025-06-10 06:11:41	SALES_PROD_WH	RESUME_WAREHOUSE	WAREHOUSE_AUTORESUME

SUMMARY ^ 125,247 rows – 5 columns

Key Takeaways:

Most recent events could be used to debug issues or figure out any automation misfires.

Data v Query History Analysis v Storage Insights v Events Insights v Credits Insights v Warehouse Efficiency Summary v

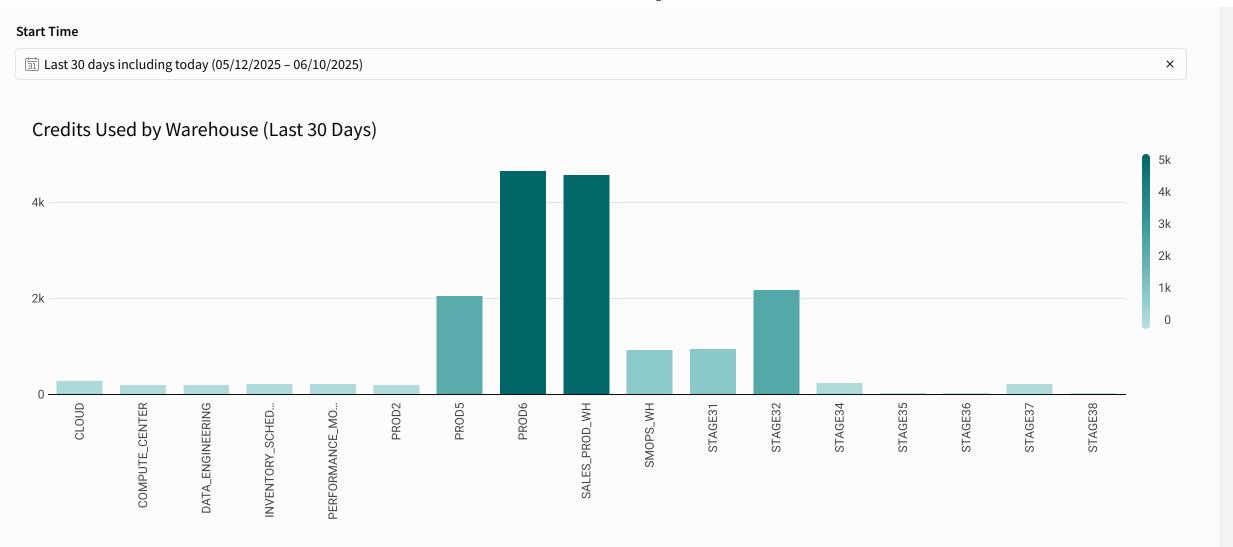
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Analysis Data sigma **Credits Consumption & Cost-Efficiency Dashboard KPI: Sum of Credits Used** Sum of Credits Used Compute Sum of Credits Used Cloud Services 5,458.46 5,449.25 9.22 √ 71.5% Jun vs May, 2025 √ 71.5% Jun vs May, 2025 √ 71.5% Jun vs May, 2025 20 10k 10k 2024-10 2024-12 2025-04 2024-06 2025-04 2024-06 2024-10

Key Takeaways: Shows the amount of credits we spent in the last 30 days, and a % change vs. the prior period. It can be used to indicate the value of our recent sizing adjustments or other implemented changes.

Also shows credits used for compute and cloud services separately to create showback or chargeback models to individual teams, enforcing financial accountability. (Used by FinOps teams)

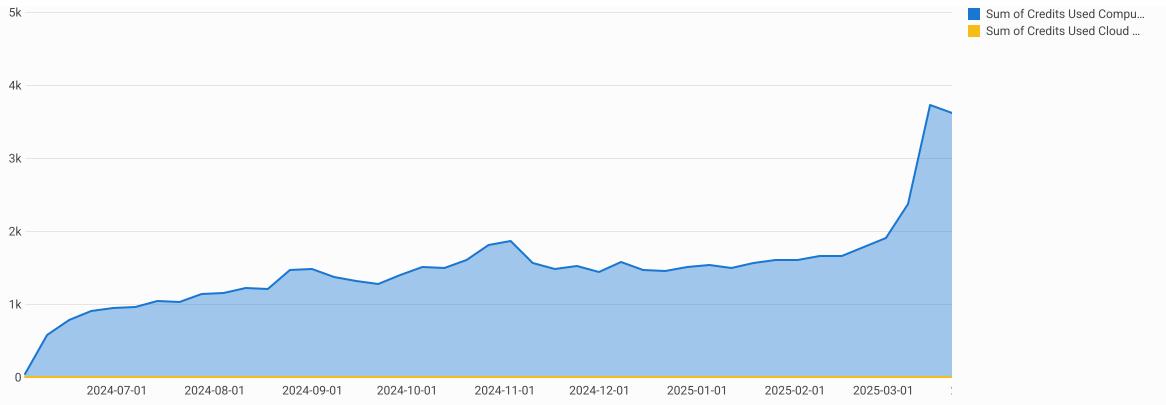
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Key Takeaways:

PROD6 and **SALES_PROD_WH** consumed **4491**\$ and **4408**\$ respectively, which means they are our top cost drivers. We should evaluate their auto-suspend settings to reduce idle spend.

Credits over Time



Key Takeaways:

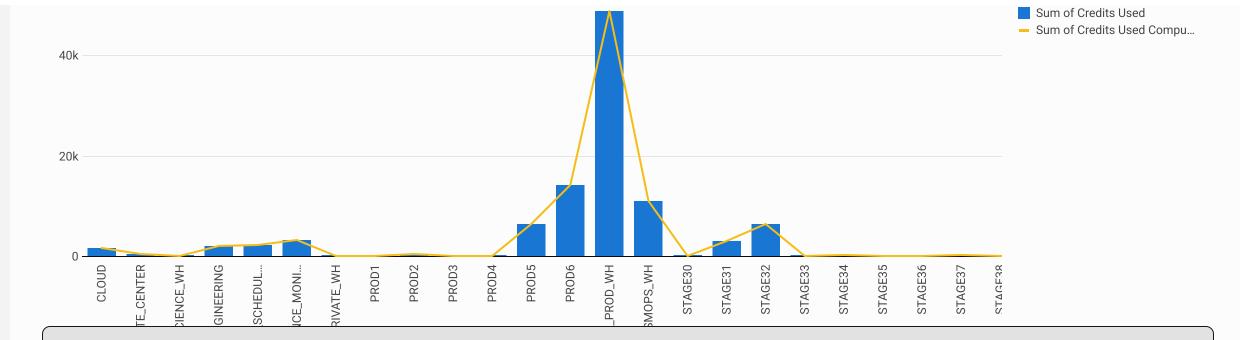
Peak usage weeks: A week that shows a 20% spike in compute spend—might correlate to batch jobs.

Shift in mix: Cloud-service credits rose from 15% to 25% of total spend, suggesting increased use of external functions or data transfers.

Downward trend: After a mid-period peak, overall credits trended down 10%, indicating recent cost-optimization measures are working.

Sum of Credits Used and Sum of Credits Used Compute by Warehouse Name

6/10/25, 3:19 PM Sigma



Key Takeaways:

Heavy compute users: Warehouses that consume almost all of total credits used for compute suggests CPU-intensive workloads.

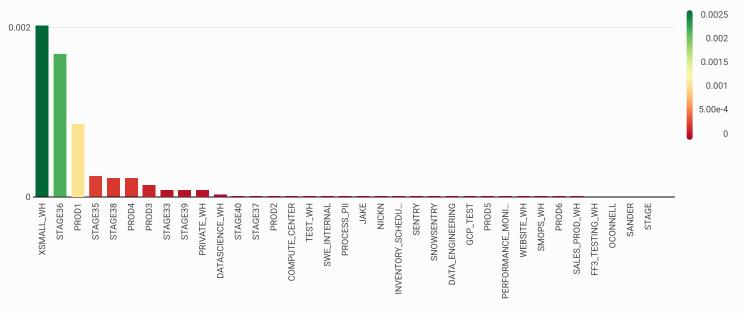
Balanced candidates: Warehouses which show roughly 1:1 ratios in compute and cloud credits are good benchmarks for optimal sizing.

Data v Query History Analysis v Storage Insights v Events Insights v Credits Insights v Warehouse Efficiency Summary v

sigma

Warehouse Efficiency & Cost-Performance Dashboard

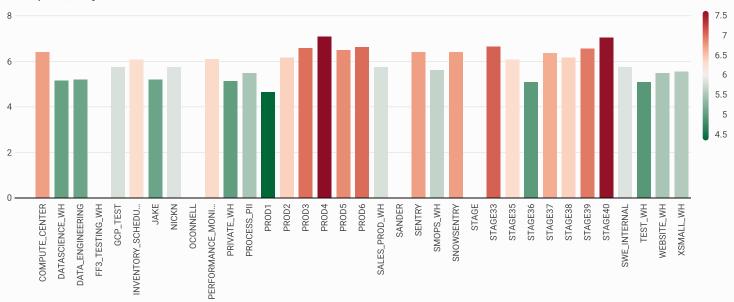
Queries per Compute Credit



Key Takeaways:

Warehouses at the top deliver the most queries per compute credit, these are our cost-efficiency benchmarks. Those at the bottom likely need resizing or query optimization.

Cost per Query



Key Takeaways:

High-cost outliers: Warehouses at the top are spending significantly more credits per query and these can becandidates for query optimization or down-sizing.

Low-cost benchmarks: Warehouses at the bottom deliver queries at a much lower cost, which are good reference configurations to replicate.