# Understanding Fabric Capacities

Benni De Jagere

Slides

CAT

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REBTECH

DATAmasterminds

Measure Killer

Snabbfoting

TIMEXTENDER

Telia

Microsoft

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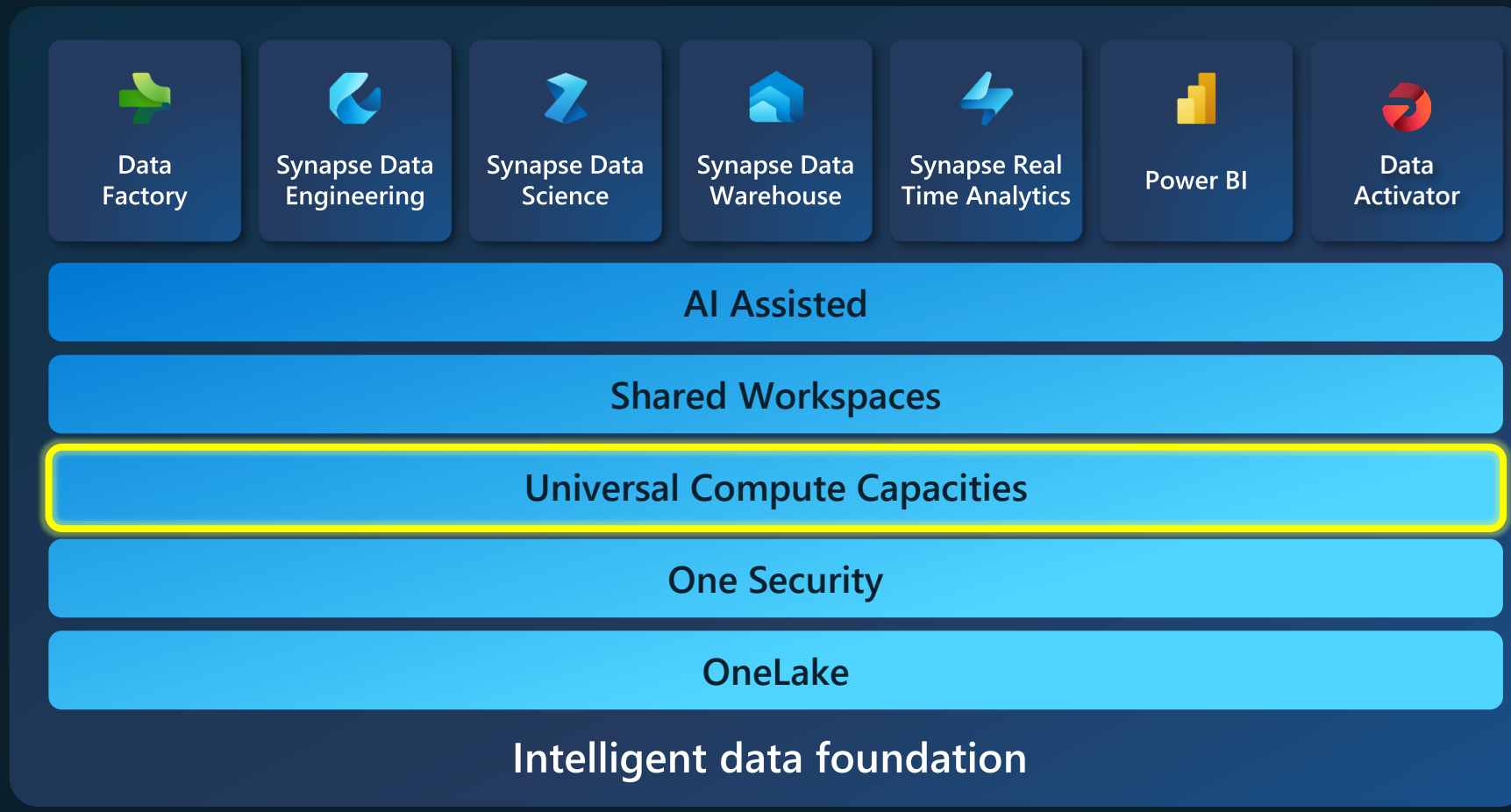
# Fabric Capacities Introduction

# Microsoft Fabric




**Single…**

Onboarding and trials
Sign-on
Navigation model
UX model
Workspace organization
Collaboration experience
Data Lake
Storage format
Data copy for all engines
Security model
CI/CD
Monitoring hub
Governance & Capacity Metrics
Data Hub

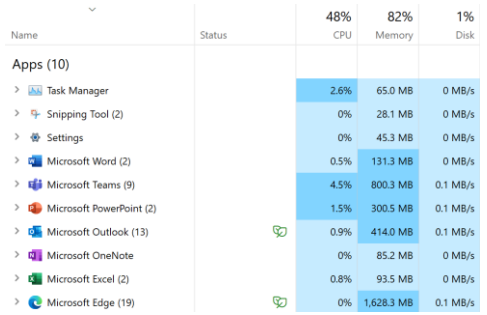
# Capacities are to Fabric what CPUs are to PCs

**Personal Computing**

When you purchase a PC you choose the number of CPU cores. The more CPU cores the more load the PC can handle.

The CPU cores are dynamically shared across all applications with no need to pre-allocate by app.

The total consumption of the CPU across all the apps cannot exceed the number of cores. CPU overload causes a slowdown.

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**Fabric Capacities**

In Fabric, you provision a Capacity with a number of "capacity units". The more capacity units provisioned, the more compute load handled.

Unlike the PC, capacity units can be scaled up or down as needed.

The capacity units are dynamically shared across all the Fabric workloads, with no pre-allocation necessary.

A single capacity can simultaneously drive BI, DW, Spark, ML and every other compute engine in Fabric

The total consumption of the capacity across all the workloads cannot exceed the capacity units provisioned.

Overloading the capacity will throttle it (slow down).

Auto scale can dynamically increase the available compute units avoiding the slowdown.

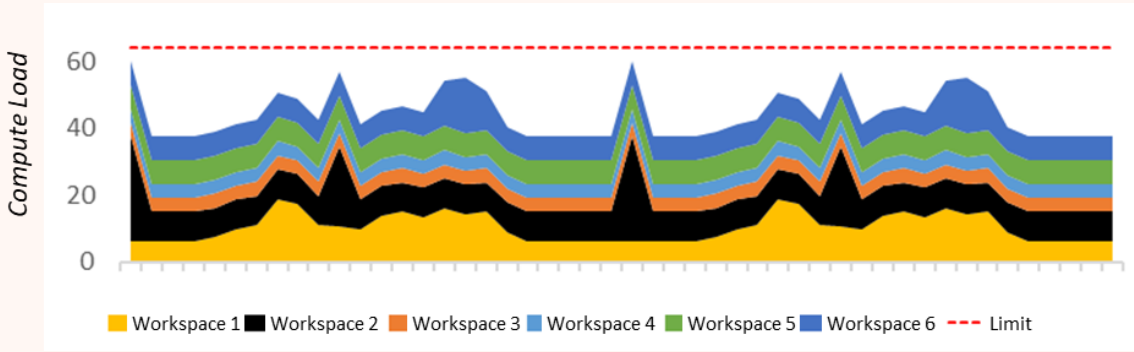
# Capacities are a shared resource

**Shared across workloads**
A single capacity is providing the compute power for all Fabric workloads.

There is no need to allocate compute for each workload separately.

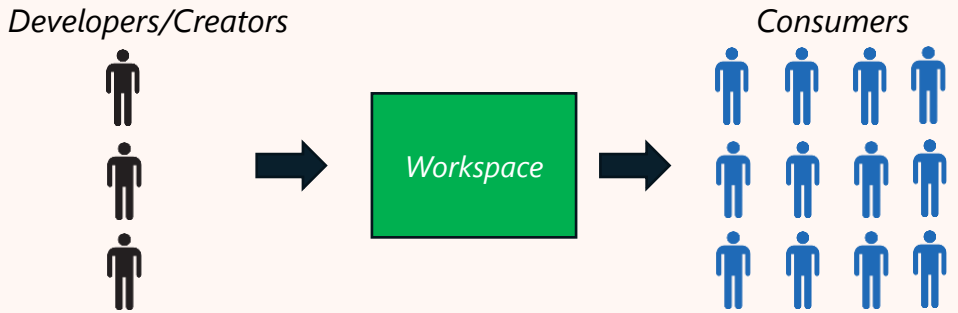

**Shared Across Projects**
A single capacity typically supports dozens of separate projects simultaneously, each managed in its own workspace.

It is rare to have a capacity dedicated to a single project




**Shared across users**
For each project, many developers will share a workspace where collaborative development and consumption at scale is managed.

Each creator can provision any artifact and run any job without the need for any pre-approval or planning

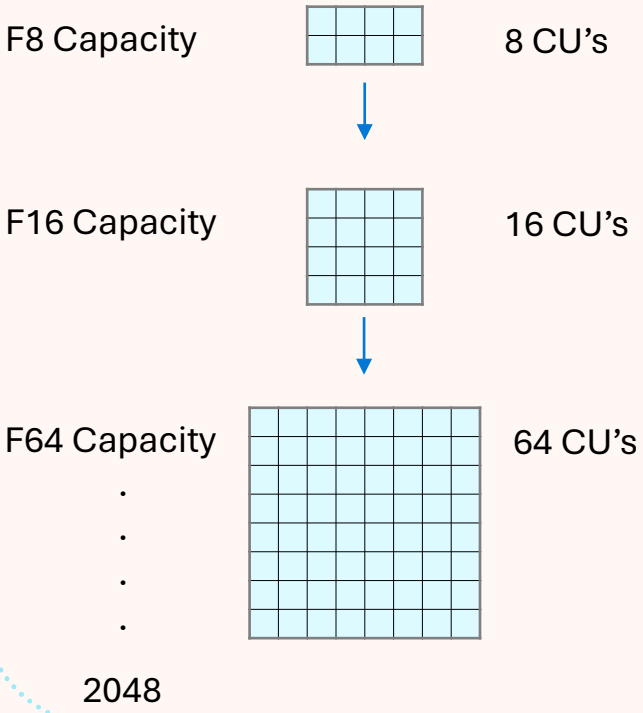


# Capacities are flexible building blocks for growth

Capacities can be configured in endless ways to meet scale, usage and governance requirements while tuning to minimize TCO and performance goals

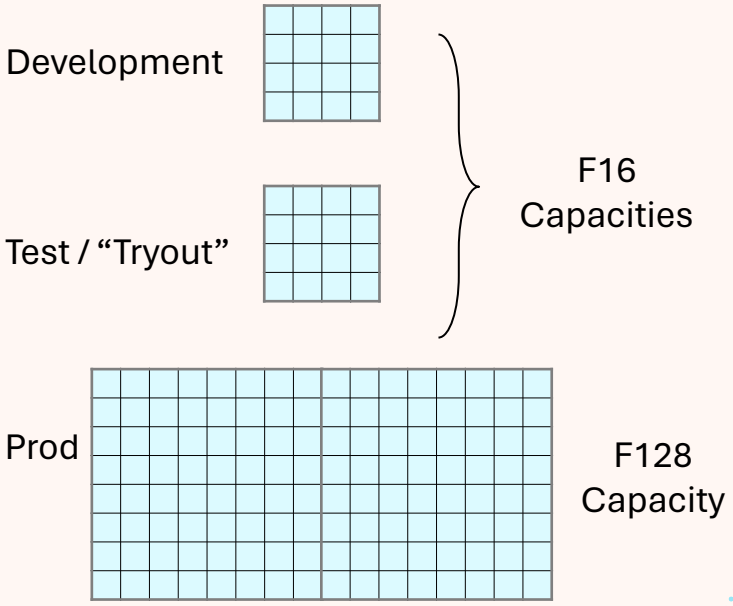
## Scale Vertically

Increased capacity size provides more throughput



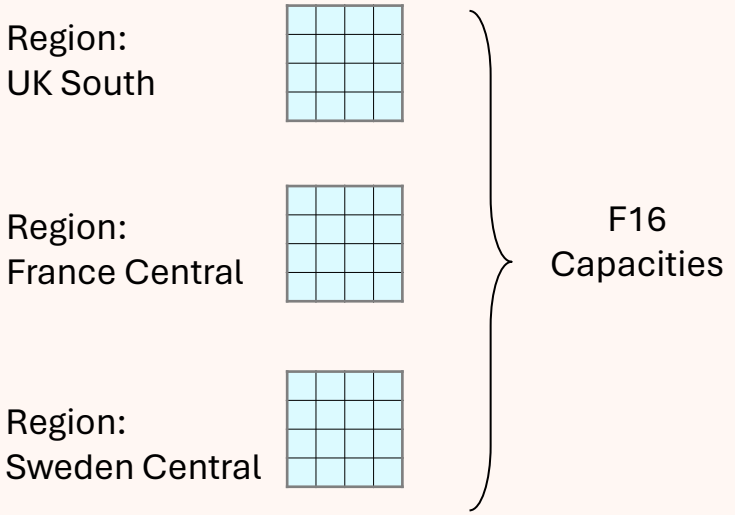

## Scale Horizontally

Scale horizontally using the benefits of modular design for hardened isolation and governance




## Regional Availability

Use different capacities for different regions to support GDPR / Data residency requirements



# Provisioning and Deploying Capacities

## Purchased in Azure

- **Purchased** either as a PAYG or RI resource
- **Provisioned with a certain amount of compute** units, analogous to CPU cores.
- The **more capacity units are provisioned, the more load** the capacity can support
  - Multiply SKU size by 30s to match platform evaluation in metrics app
- Capacities are **priced at a fixed hourly rate**, based on capacity units provisioned
- The RI commitment (1-year reserved instance) enjoys a **40% discount**

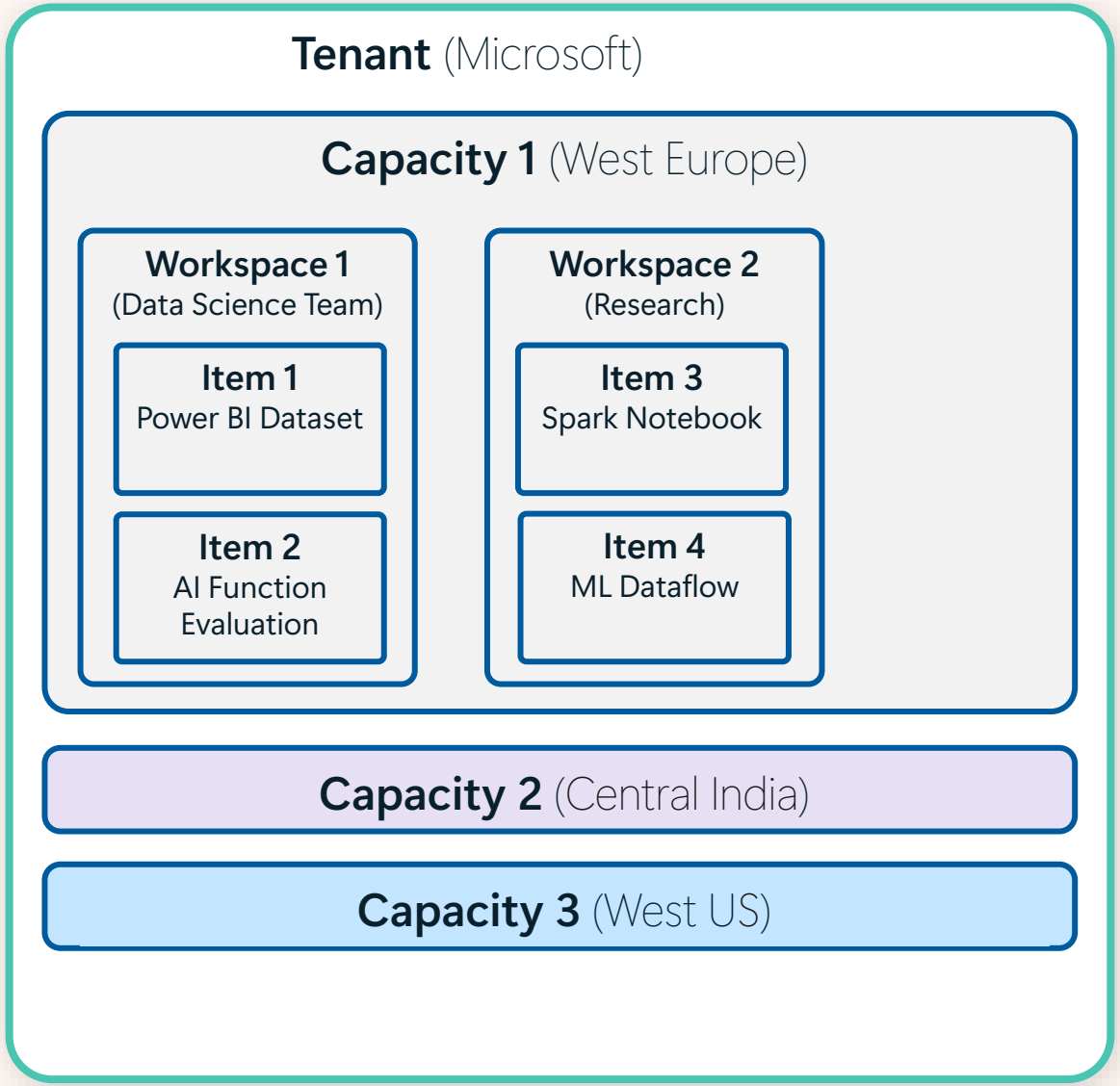
Universal Compute Capacities SKU Sizing

| SKU | Capacity Units (CU) | CU's (per 30s) | Power BI SKU | Power BI V-cores |
|---|---|---|---|---|
| F2 | 2 | 60 | - | 0.25 |
| F4 | 4 | 120 | - | 0.5 |
| F8 | 8 | 240 | A1 | 1 |
| F16 | 16 | 480 | A2 | 2 |
| F32 | 32 | 960 | A3 | 4 |
| F64 | 64 | 1920 | P1 | 8 |
| F128 | 128 | 3840 | P2 | 16 |
| F256 | 256 | 7680 | P3 | 32 |
| F512 | 512 | 15360 | P4 | 64 |
| F1024 | 1024 | 30720 | P5 | 128 |
| F2048 | 2048 | 61440 | - | 256 |

# Provisioning and Deploying Capacities

## Deployed to Regions

- Each capacity **resides in a specific region of the buyers' choice** where both the data & compute reside
- **Workspaces are assigned to a capacity** that provides the compute and storage for all the workspace artifacts
- Multiple capacities can be purchased, deployed and managed by **different owners** residing in a single tenant allowing each business unit to pay for their own consumption



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# Bursting and Smoothing

# Smoothing intro and benefits

| | |
|---|---|
| Load stabilization | Smoothing helps capacities self-stabilize by flattening large spikey loads into a smooth load profile, eliminating temporal spikes |
| Eliminates Scheduling contention | Large/scheduled Jobs usage (not execution) are smoothed over 24 hours, eliminating the need to decide the timing and order of job execution<br><br>*Performance now, pay later. Demo to follow.* |
| Bad actor protection | Interactive operations smoothed over several minutes, preventing a single user with a very demanding query from hogging the entire capacity |

# What is Bursting?

## Job acceleration

Bursting provides extra compute resources to jobs and queries to accelerate their completion

## Go beyond

The extra resources of bursting allow jobs to **utilize far more resources than "face value"**

Instead of running a job on 64 CU and completing in 60 seconds, bursting could use 256 CUs to complete the job in 15 seconds.

Same amount of work, just completed faster

## No hassle, No overload

**Bursting is automatic** when the system reasons it can accelerate the job by applying extra resources. No settings are required.

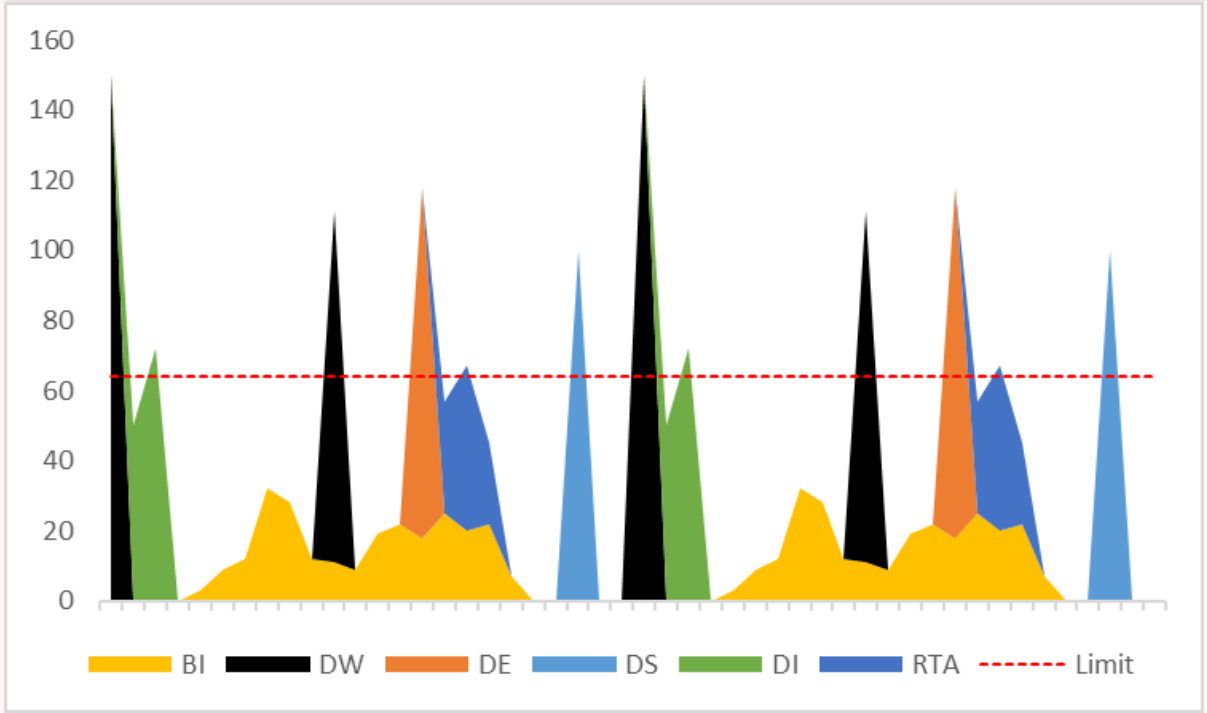
**Bursting prevents an overload** as the *smoothing* mechanism will always flatten the resource burst

# Bursting and smoothing | before and after

Looking at an example of a 64 CU capacity, running multiple workloads over a couple of days…

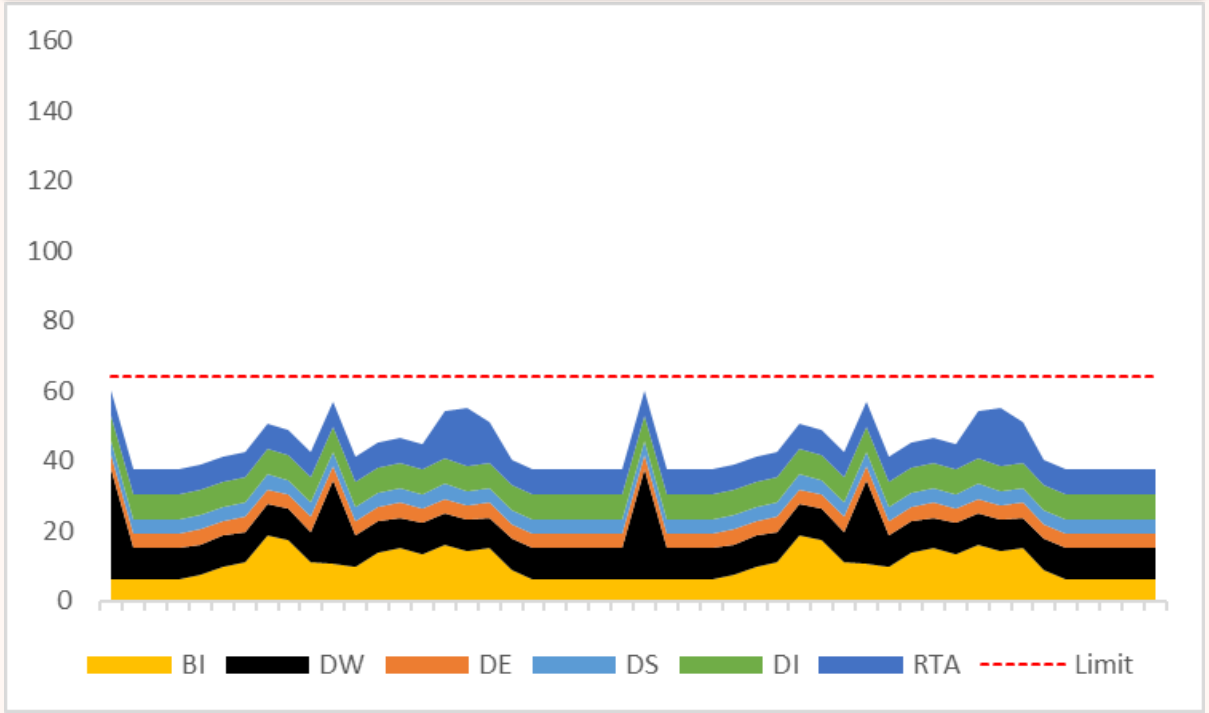
## Before Smoothing

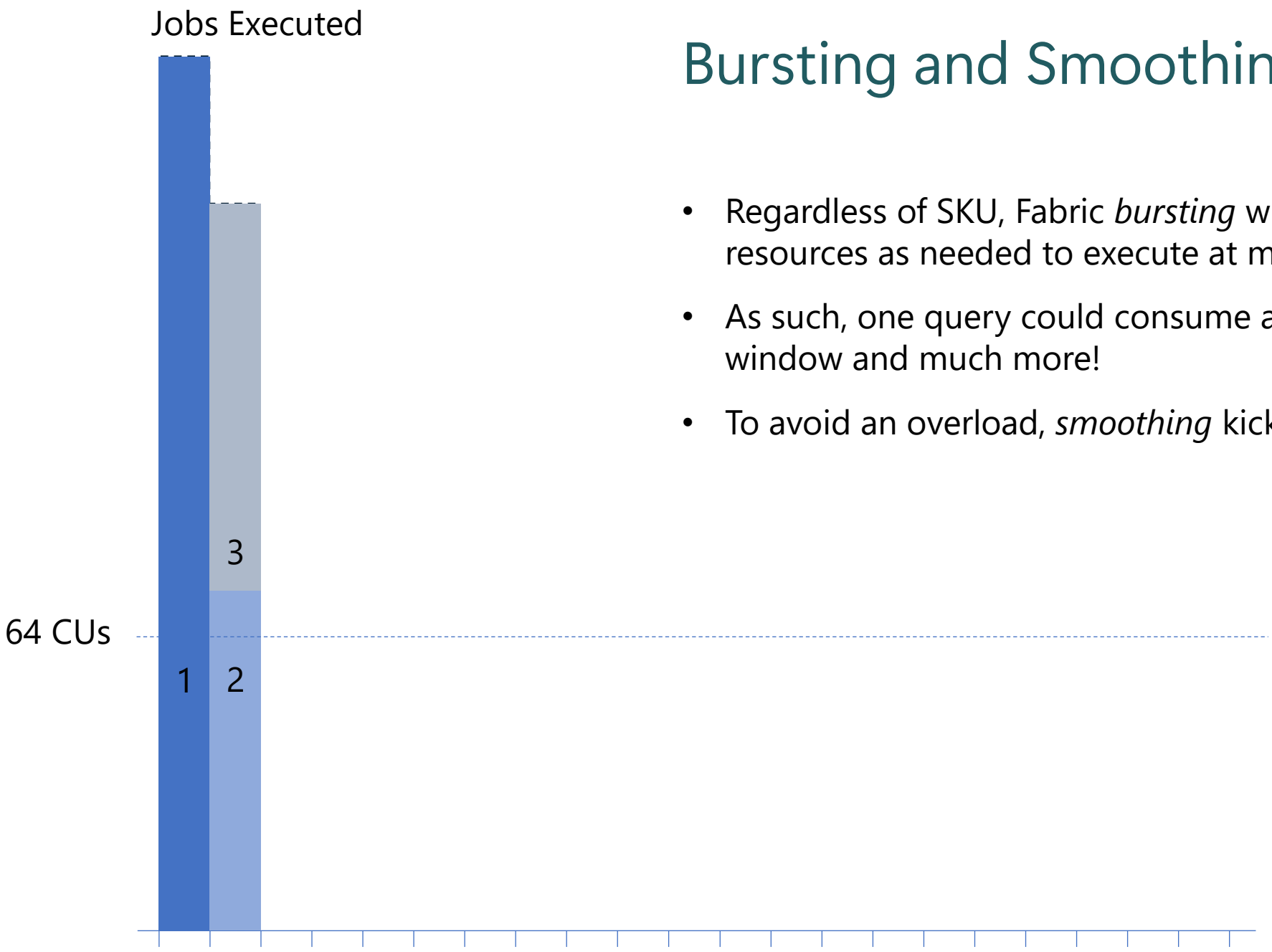
- Actual load as executed on the capacity before smoothing
- *Bursting* accelerates jobs execution by resource boosting
- The capacity could be overloaded 25% of the time
- Some of the overloads are more than 2x the limit
- There are periods of no/low usage




## After Smoothing

- Shows the reported load (not runtime execution) against the capacity limits
- There is NO overload, and consumption is more stable
- The smoothing of usage fills in gaps

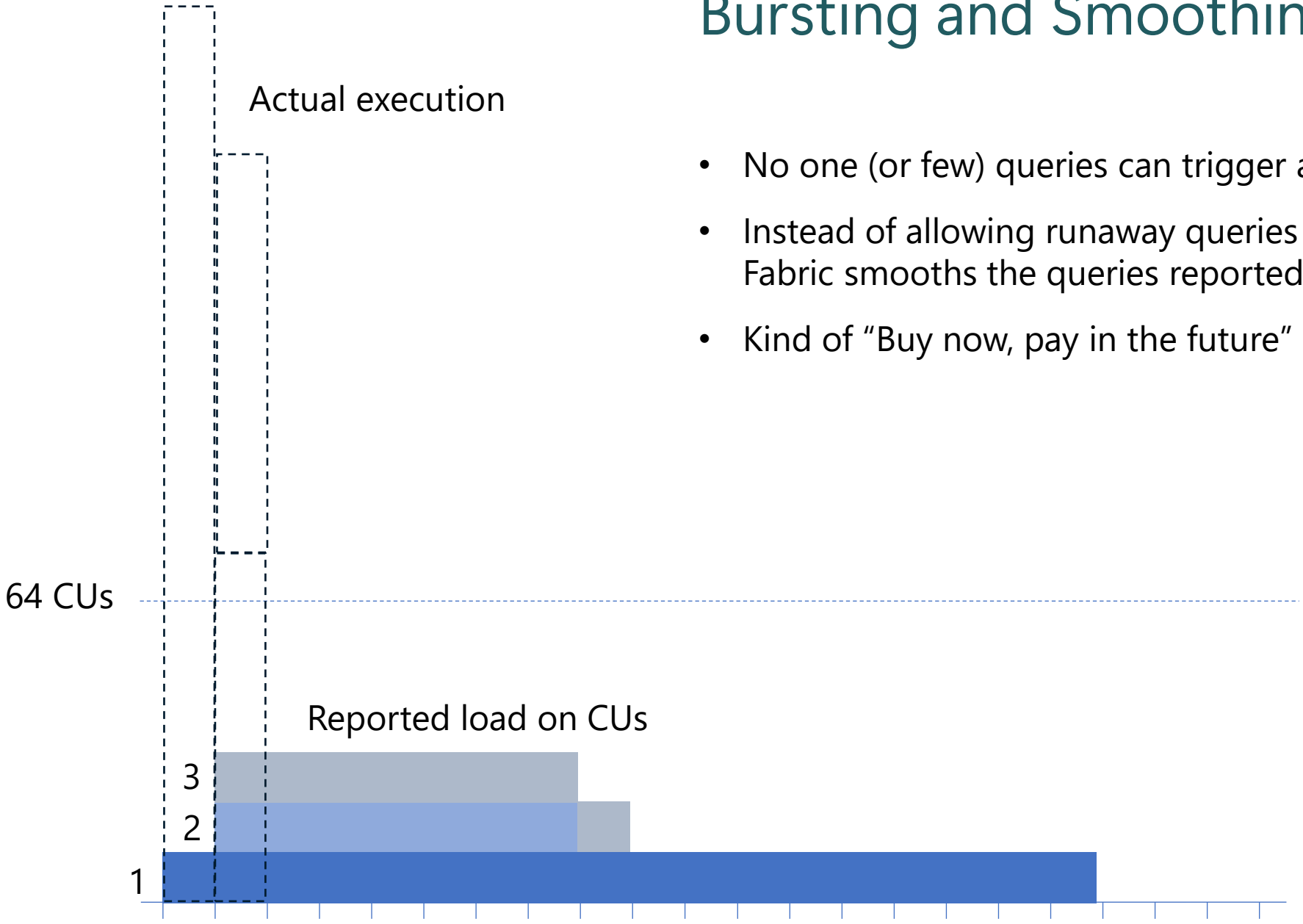





# Bursting and Smoothing

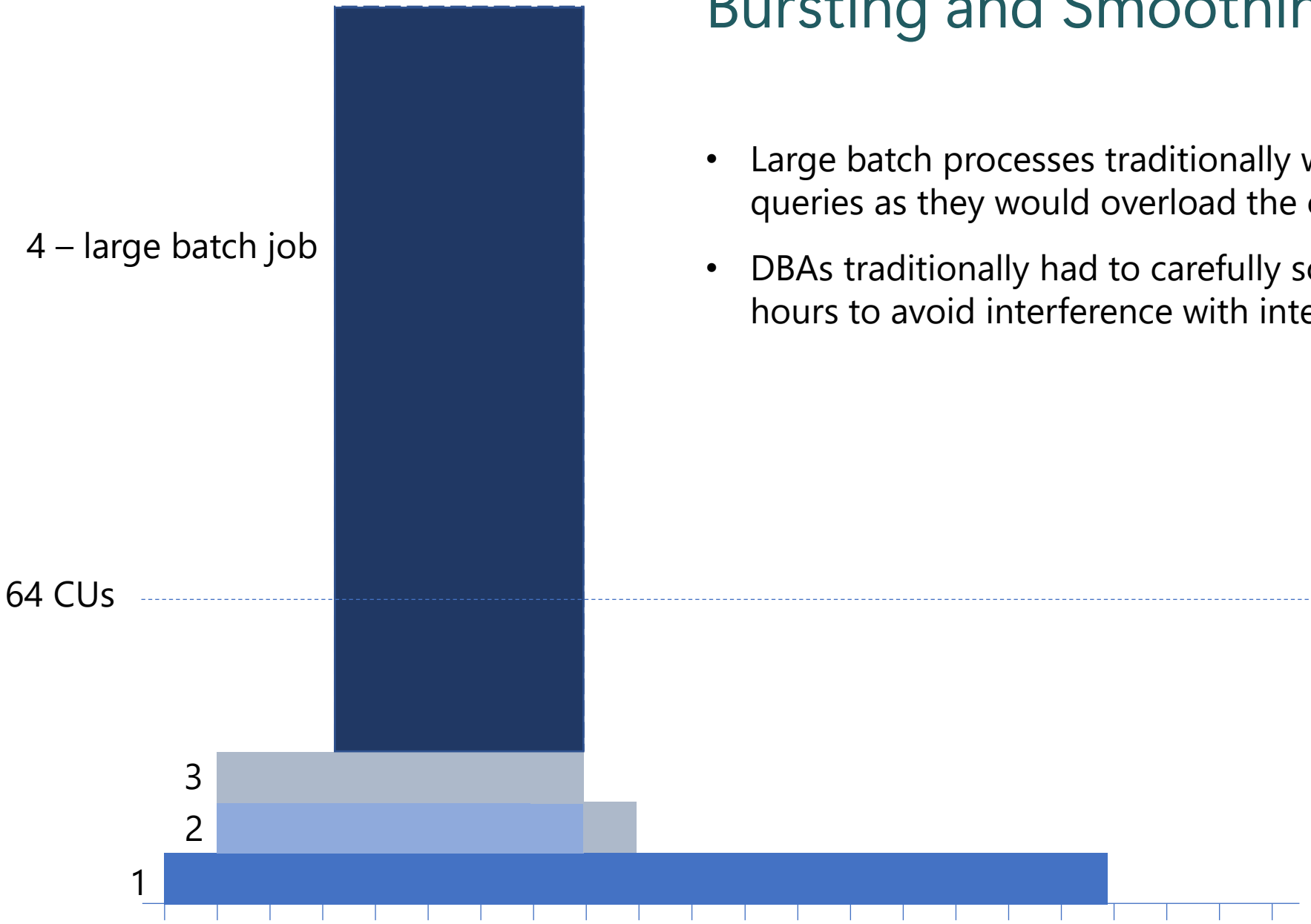
- Regardless of SKU, Fabric *bursting* will automatically allocate resources as needed to execute at maximum performance
- As such, one query could consume all the quota of a single time window and much more!
- To avoid an overload, *smoothing* kicks in

# Bursting and Smoothing




- No one (or few) queries can trigger an overload
- Instead of allowing runaway queries to create a local overload, Fabric smooths the queries reported usage to future time windows
- Kind of “Buy now, pay in the future” installment plan

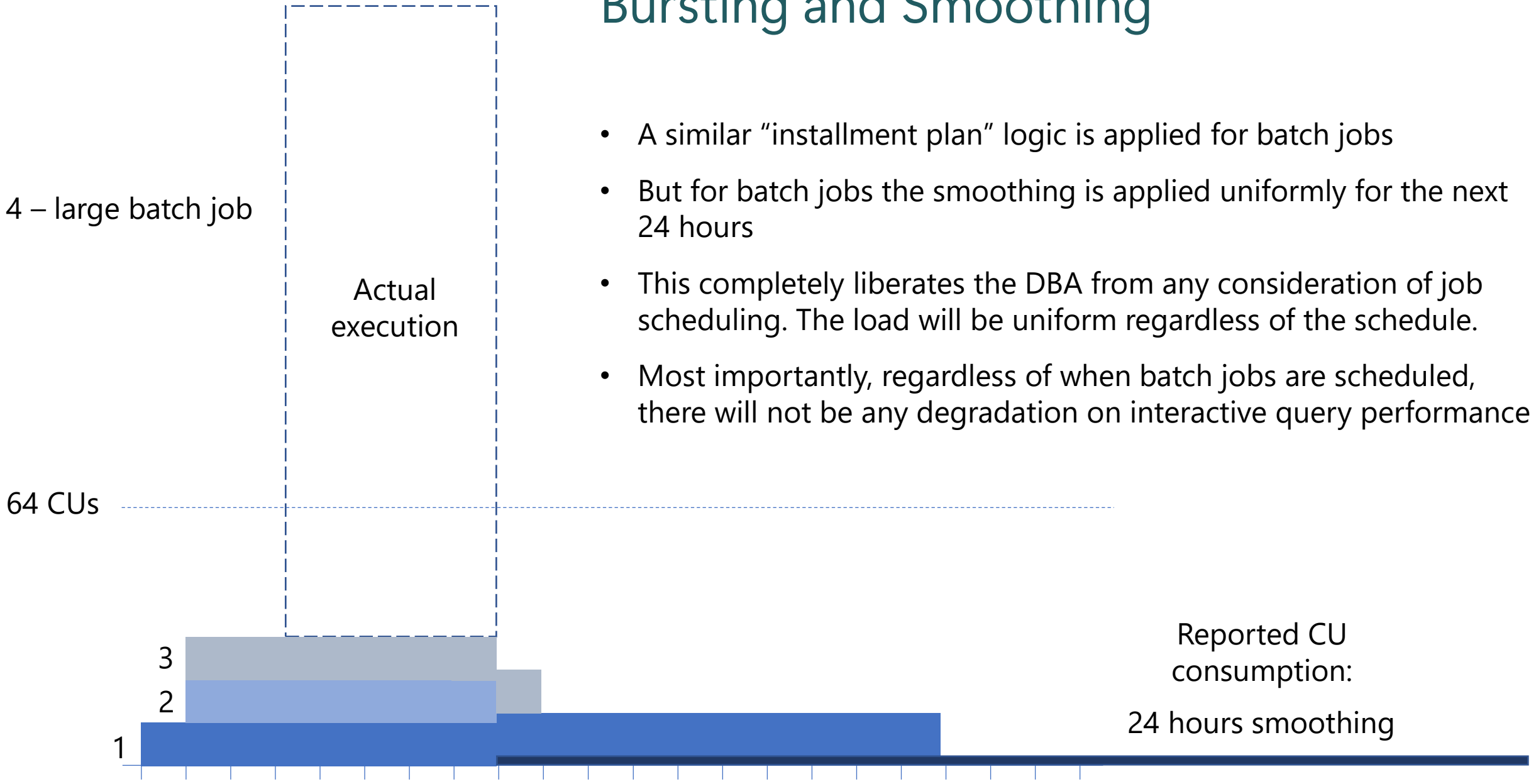
# Bursting and Smoothing




- Large batch processes traditionally were a threat to interactive queries as they would overload the compute resource
- DBAs traditionally had to carefully schedule these jobs to off-hours to avoid interference with interactive user experiences

# Bursting and Smoothing

- A similar “installment plan” logic is applied for batch jobs
- But for batch jobs the smoothing is applied uniformly for the next 24 hours
- This completely liberates the DBA from any consideration of job scheduling. The load will be uniform regardless of the schedule.
- Most importantly, regardless of when batch jobs are scheduled, there will not be any degradation on interactive query performance



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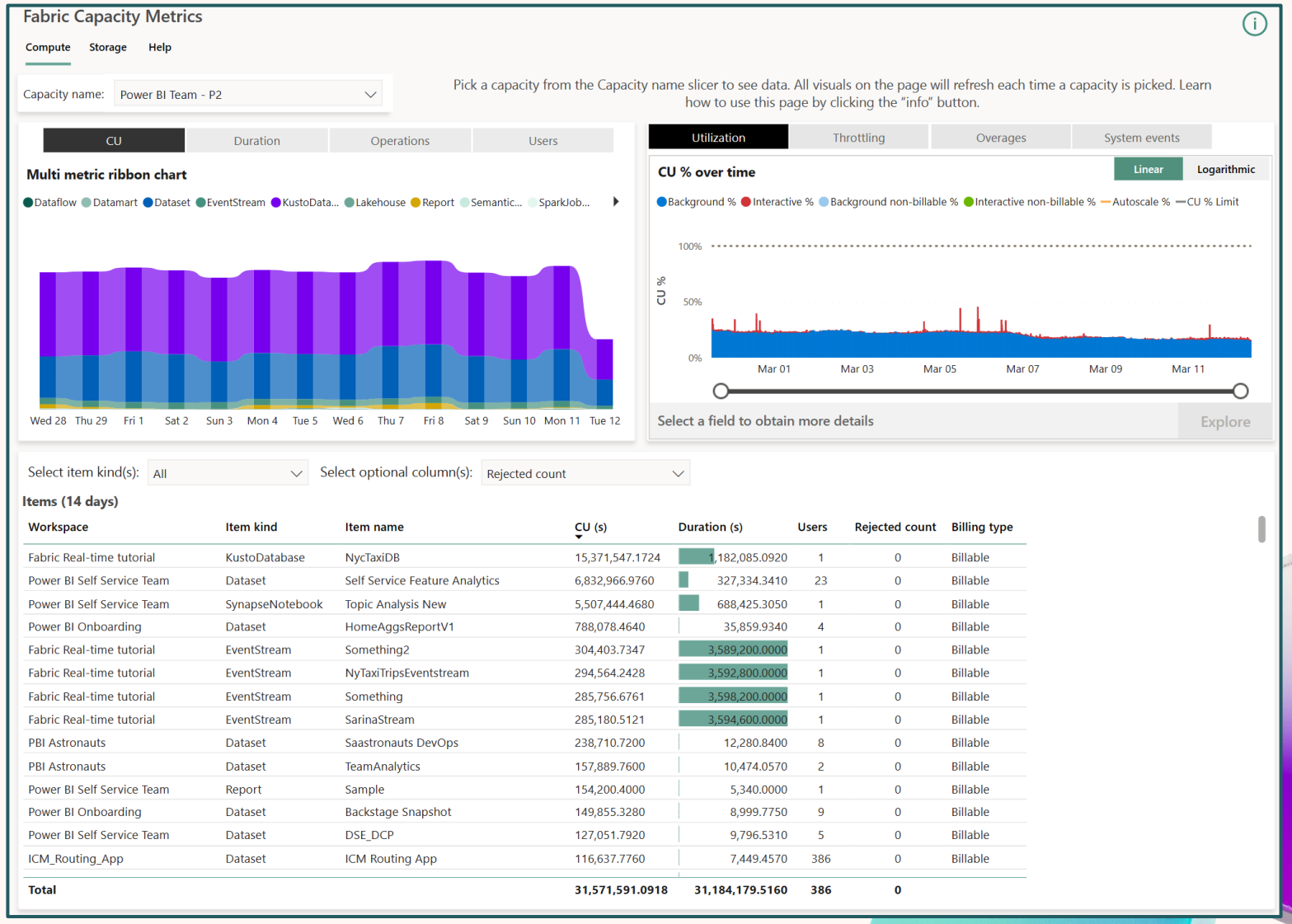
# Monitoring with Capacity Metrics

# Capacity Metrics

## Monitor Capacities and Plan capacity scale-up with confidence

- Tenant wide visibility into capacity usage for all Fabric experiences
- Identify resource usage trends and their impact to autoscale & throttling
- View preview workload usage alongside production workloads to make data-driven capacity sizing decisions



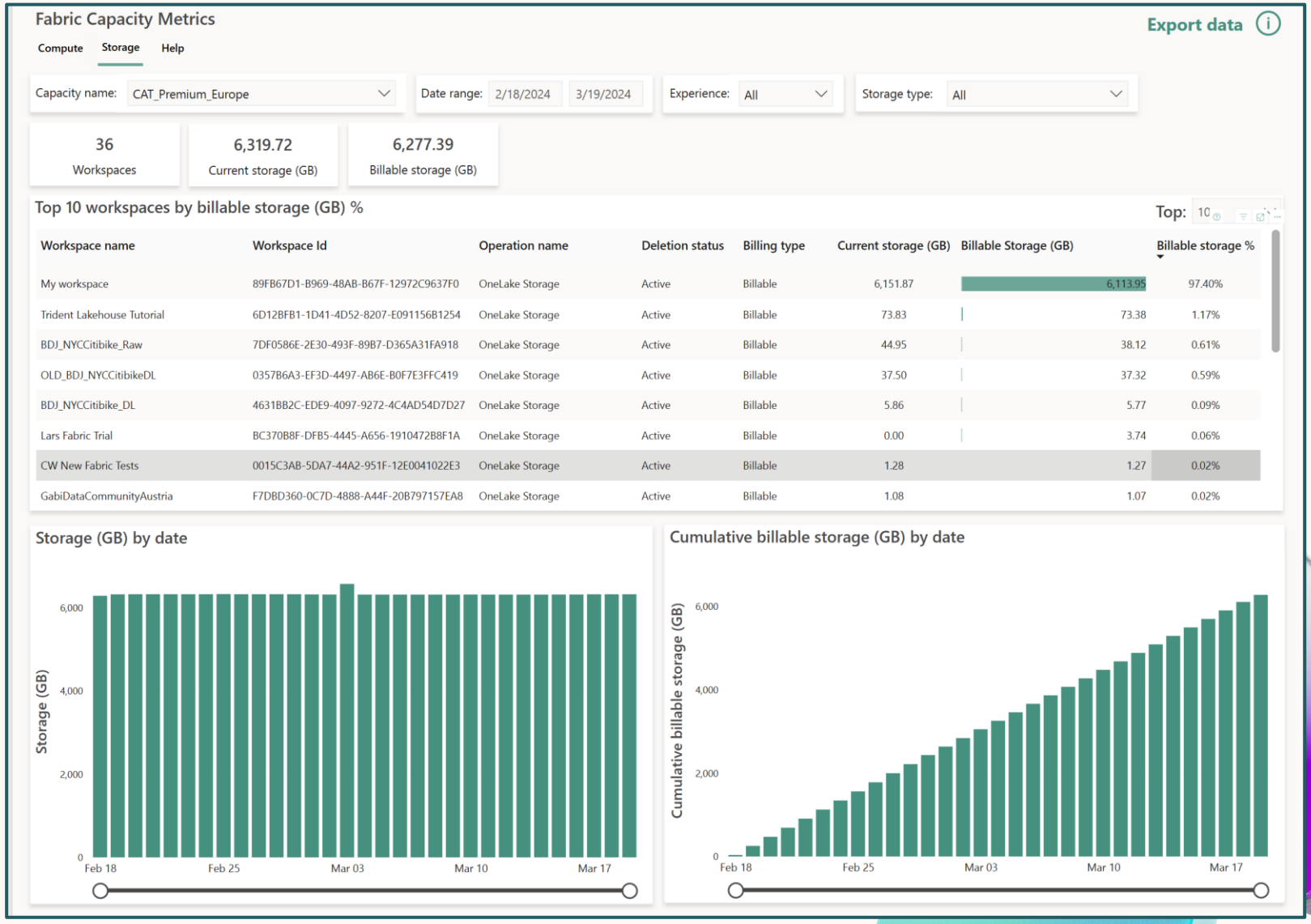

| Workspace | Item kind | Item name | CU (s) | Duration (s) | Users | Rejected count | Billing type |
|---|---|---|---|---|---|---|---|
| Fabric Real-time tutorial | KustoDatabase | NycTaxiDB | 15,371,547.1724 | 1,182,085.0920 | 1 | 0 | Billable |
| Power BI Self Service Team | Dataset | Self Service Feature Analytics | 6,832,966.9760 | 327,334.3410 | 23 | 0 | Billable |
| Power BI Self Service Team | SynapseNotebook | Topic Analysis New | 5,507,444.4680 | 688,425.3050 | 1 | 0 | Billable |
| Power BI Onboarding | Dataset | HomeAggsReportV1 | 788,078.4640 | 35,859.9340 | 4 | 0 | Billable |
| Fabric Real-time tutorial | EventStream | Something2 | 304,403.7347 | 3,589,200.0000 | 1 | 0 | Billable |
| Fabric Real-time tutorial | EventStream | NyTaxiTripsEventstream | 294,564.2428 | 3,592,800.0000 | 1 | 0 | Billable |
| Fabric Real-time tutorial | EventStream | Something | 285,756.6761 | 3,598,200.0000 | 1 | 0 | Billable |
| Fabric Real-time tutorial | EventStream | SarinaStream | 285,180.5121 | 3,594,600.0000 | 1 | 0 | Billable |
| PBI Astronauts | Dataset | Saastronauts DevOps | 238,710.7200 | 12,280.8400 | 8 | 0 | Billable |
| PBI Astronauts | Dataset | TeamAnalytics | 157,889.7600 | 10,474.0570 | 2 | 0 | Billable |
| Power BI Self Service Team | Report | Sample | 154,200.4000 | 5,340.0000 | 1 | 0 | Billable |
| Power BI Onboarding | Dataset | Backstage Snapshot | 149,855.3280 | 8,999.7750 | 9 | 0 | Billable |
| Power BI Self Service Team | Dataset | DSE_DCP | 127,051.7920 | 9,796.5310 | 5 | 0 | Billable |
| ICM_Routing_App | Dataset | ICM Routing App | 116,637.7760 | 7,449.4570 | 386 | 0 | Billable |
| **Total** | | | **31,571,591.0918** | **31,184,179.5160** | **386** | **0** | |

# Capacity Metrics

## Monitor OneLake consumption

Measure the trends of workspace storage consumption against capacity limits, by day or hour

Reconcile costs with internal chargeback processes



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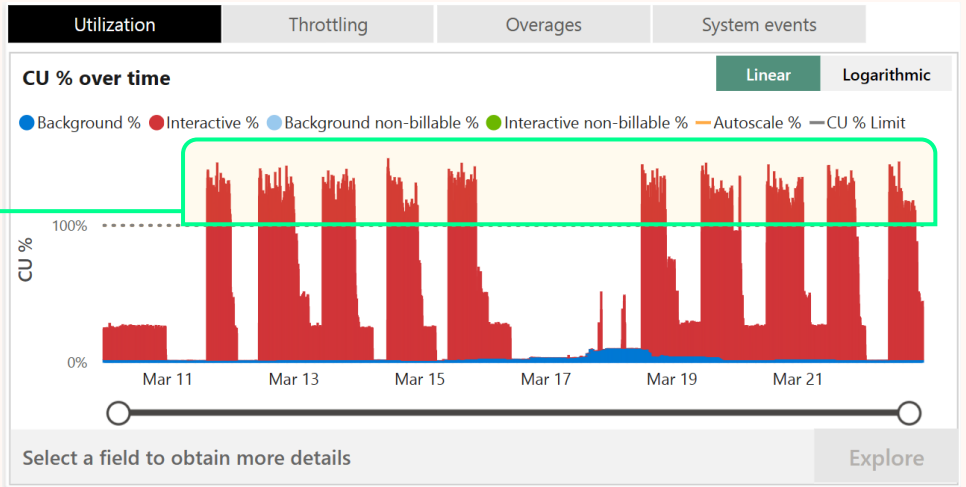
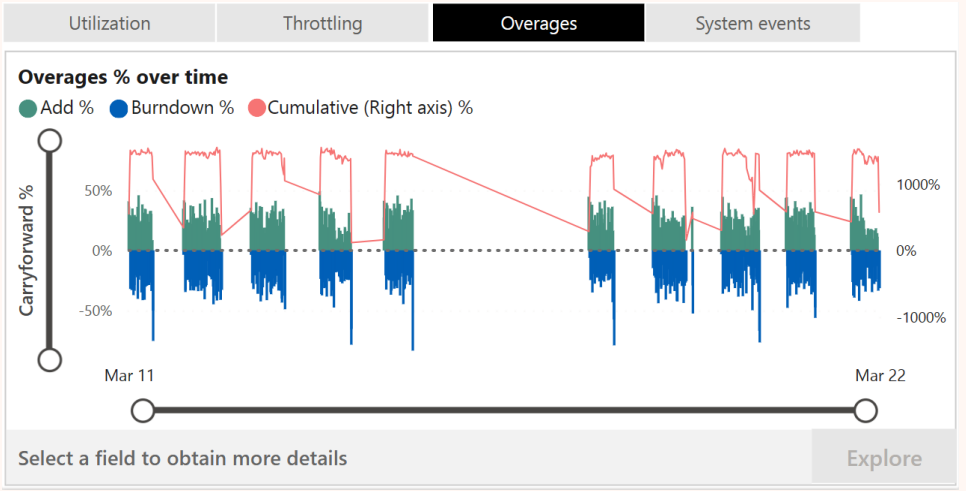
# Capacity Throttling Policies

# Throttling intro

- Throttling is the platform policy for managing consumption that exceeds throughput is provided by SKU choice
- When workloads exceed the throughput of a capacity a cumulative debt is tracked to be burned down
- Cumulative debt is used to determine throttling policies and is burned down when resources are free

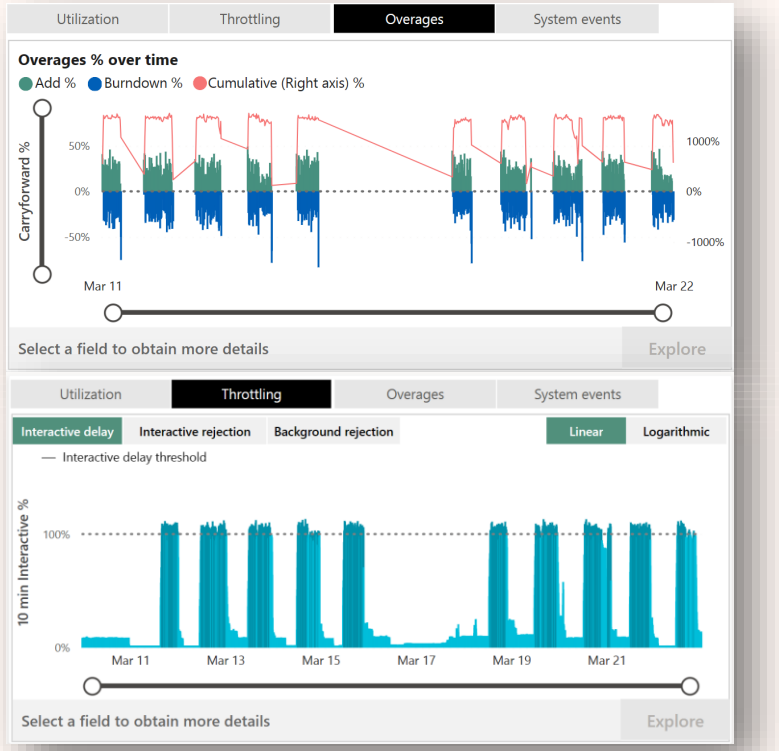



| Overage Operation | Description |
|---|---|
| Overages - Added | • Timepoint when job requests exceed the throughput of a capacity, overages are was added to the cumulative buffer to burn down.<br>• This graph simplifies identification of the optimal timepoint to load timepoint drill to analyze the user operations that contributed to an overage. |
| Overages - Burndown | • Overages being reconciled when future capacity is free to burn down |
| Overages - Cumulative | • The total amount of queued work on the capacity to be burned down in the future when the capacity is not fully utilized |



# Capacity throttling evolution for Fabric

- For Fabric, throttling policies were refined to deliver multiple benefits
  - **Reduced throttling** for capacities that only experience occasional spikes
  - **Added overage protection** – rejection policies prevent overloaded capacities from irrecoverable overload
  - **Optimizations for long-running jobs:** We're optimizing the platform for long-running jobs, so if a job exceeds capacity limits, it will run to completion and the overage will be burned down against future capacity




| Smoothed Capacity - Future Use | Platform Policy | Customer Impact |
|---|---|---|
| <= 10m | Overage Protection | Jobs can consume 10 minutes of future capacity use without throttling |
| > 10m → <= 60m | Interactive Delay | User requested interactive type jobs will be throttled |
| > 60m → <= 24h | Interactive Rejection | User requested interactive type jobs will be rejected |
| > 24h | Background Rejection | User Scheduled background jobs will be rejected from execution |

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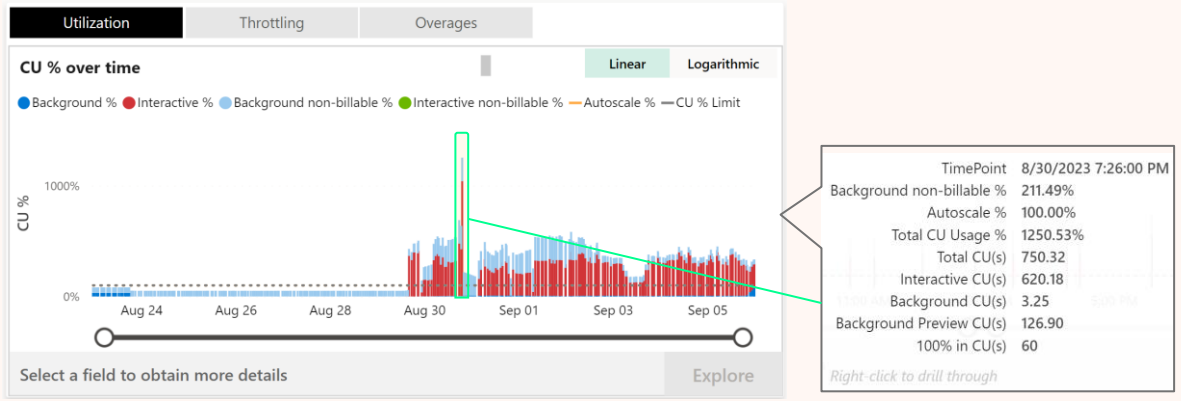
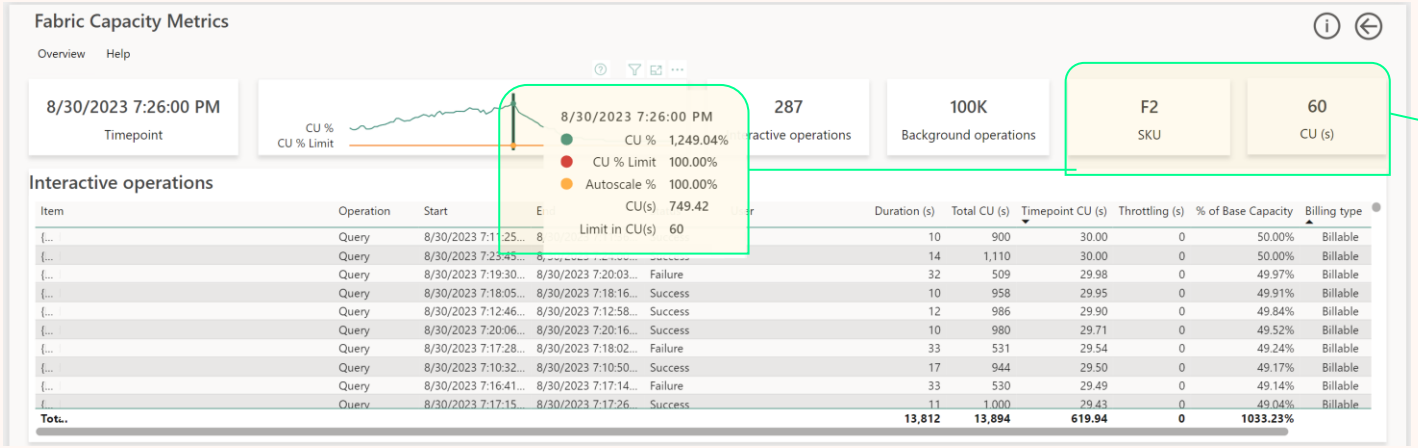
# Capacity Planning with Capacity Metrics

# Capacity planning case study - measurement

Start with a test or trial capacity to evaluate the load of specific Fabric Experiences i.e., Power BI Datasets, Spark Notebooks or a Datawarehouse




If usage is above the current capacity limits , choose the desired utilization rate to accommodate via capacity scale up




Load Capacity Metrics timepoint drill to analyze :

- Total CU's consumed : 749 CU(s)
- Capacity Size : (F2)
- CU(s) available on your capacity : 60 CU(s)

# Capacity planning case study – SKU selection

To accommodate a 749 CU(s) load the admin can purchase an F32 capacity providing 960 CU(s) of throughput

Universal Compute Capacities SKU Sizing

| SKU | Capacity Units (CU) | CU's (per 30s) | Power BI SKU | Power BI V-cores |
|---|---|---|---|---|
| F2 | 2 | 60 | - | 0.25 |
| F4 | 4 | 120 | - | 0.5 |
| F8 | 8 | 240 | A1 | 1 |
| F16 | 16 | 480 | A2 | 2 |
| F32 | 32 | 960 | A3 | 4 |
| F64 | 64 | 1920 | P1 | 8 |
| F128 | 128 | 3840 | P2 | 16 |
| F256 | 256 | 7680 | P3 | 32 |
| F512 | 512 | 15360 | P4 | 64 |
| F1024 | 1024 | 30720 | P5 | 128 |
| F2048 | 2048 | 61440 | - | 256 |

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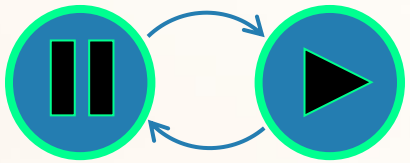
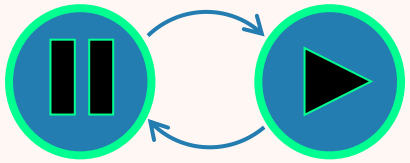
# Pausing and Resuming Capacities



# Introduction to Pausing and Resuming Capacities



## Overview and Benefits

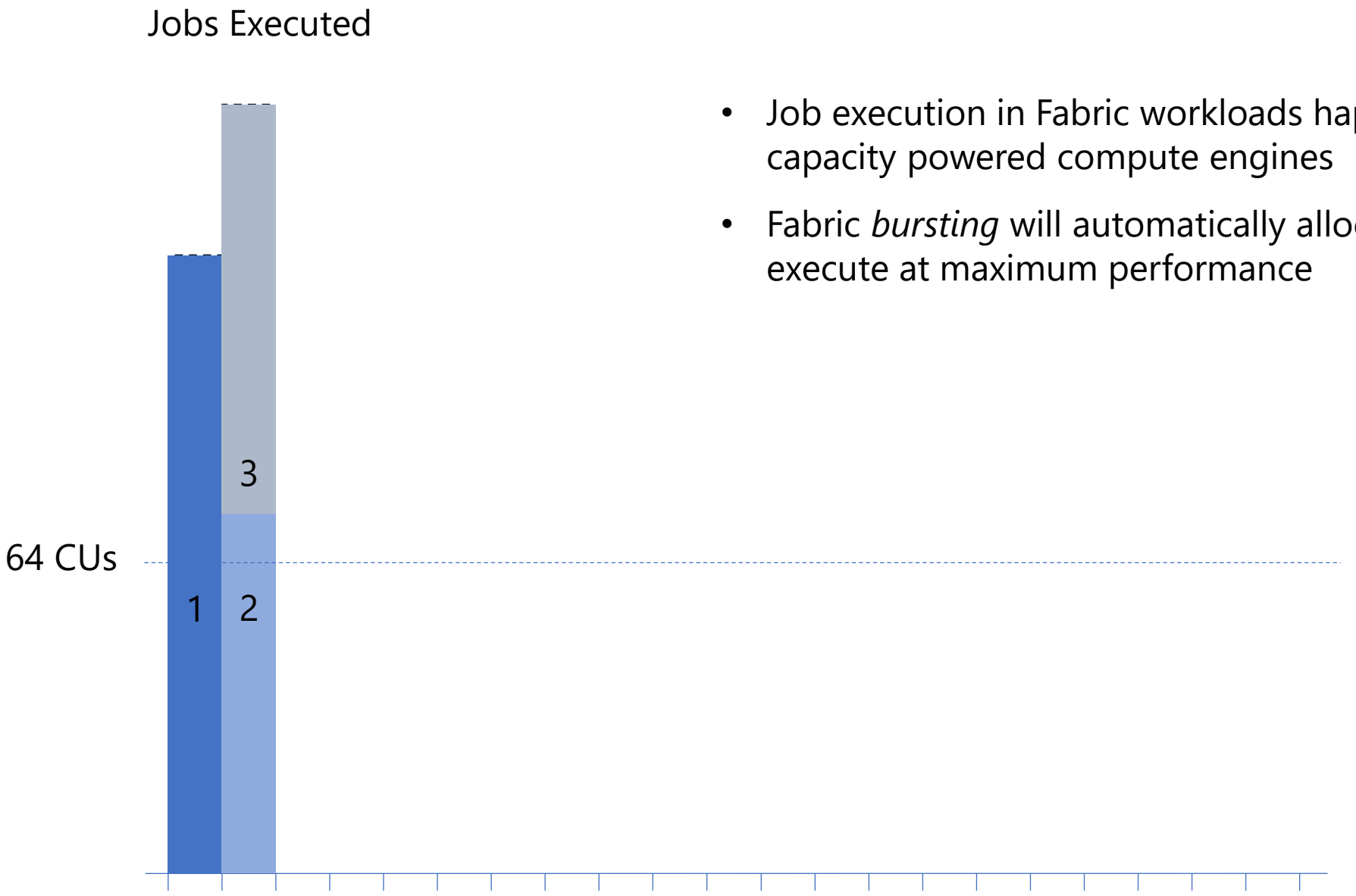
**Pause and Resume** lets you manage compute costs on F SKU capacities by suspending the execution of all workloads running on the capacity

- When a capacity administrator pauses a capacity:

| | | |
|---|---|---|
| Workloads stop execution | New requests are not run | Smoothed usage will be reconciled<br><br>(details in the next demo) |

**Note:** OneLake storage will remain active and billable while  a capacity is paused

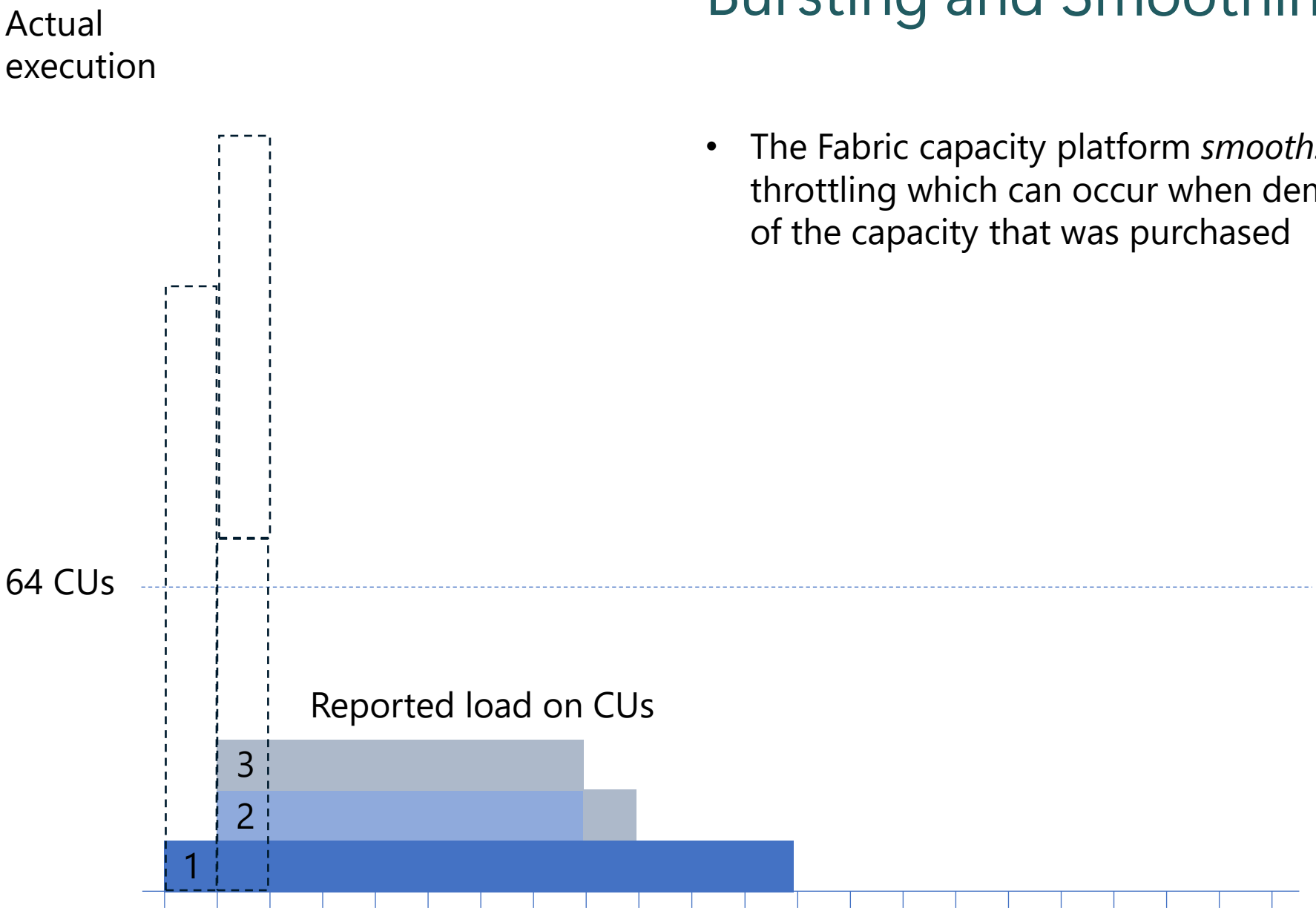
# Bursting and Smoothing




- Job execution in Fabric workloads happens on-demand via capacity powered compute engines
- Fabric *bursting* will automatically allocate resources as needed to execute at maximum performance

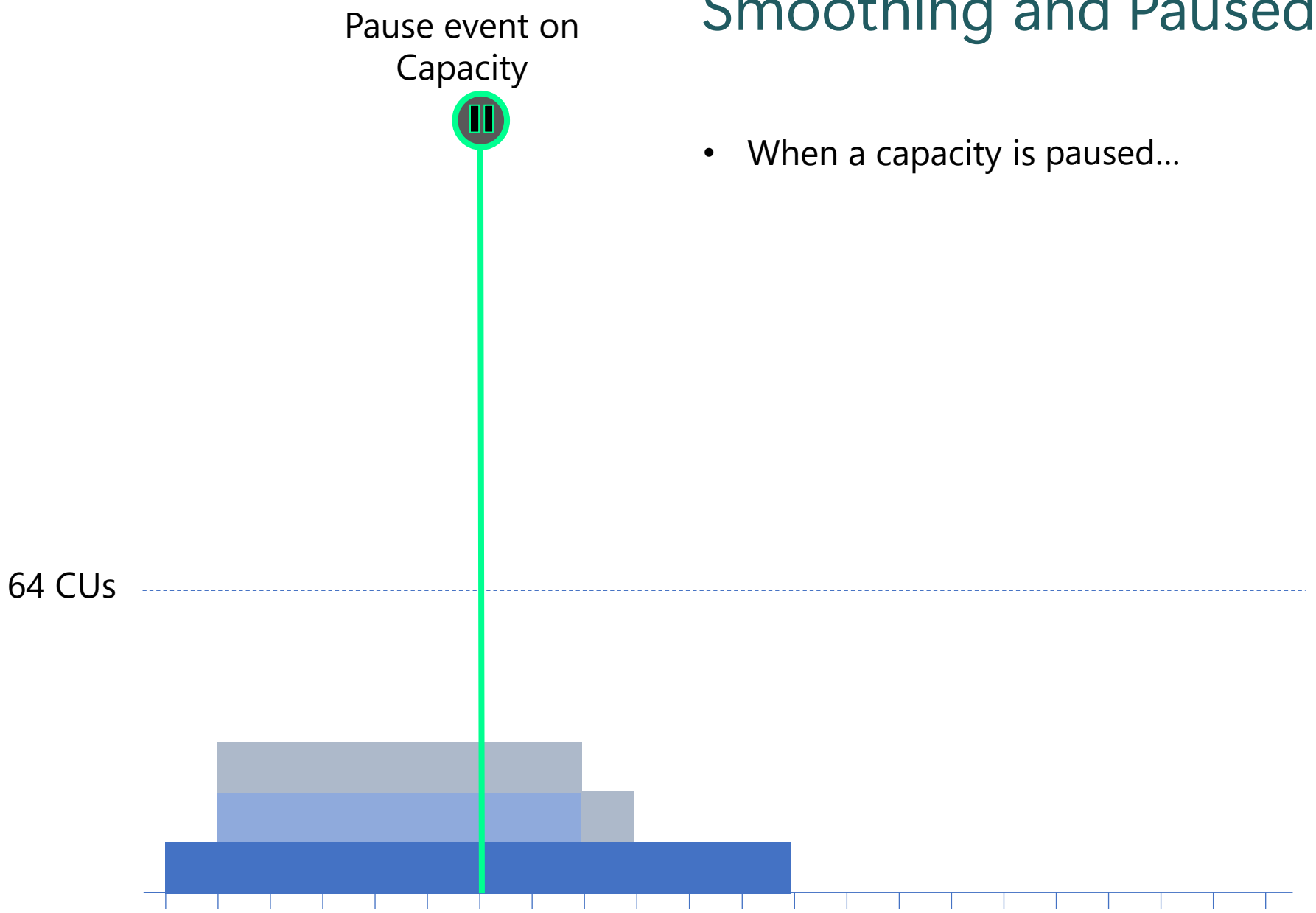
# Bursting and Smoothing

- The Fabric capacity platform *smooths* usage out to reduce throttling which can occur when demand exceeds the throughput of the capacity that was purchased

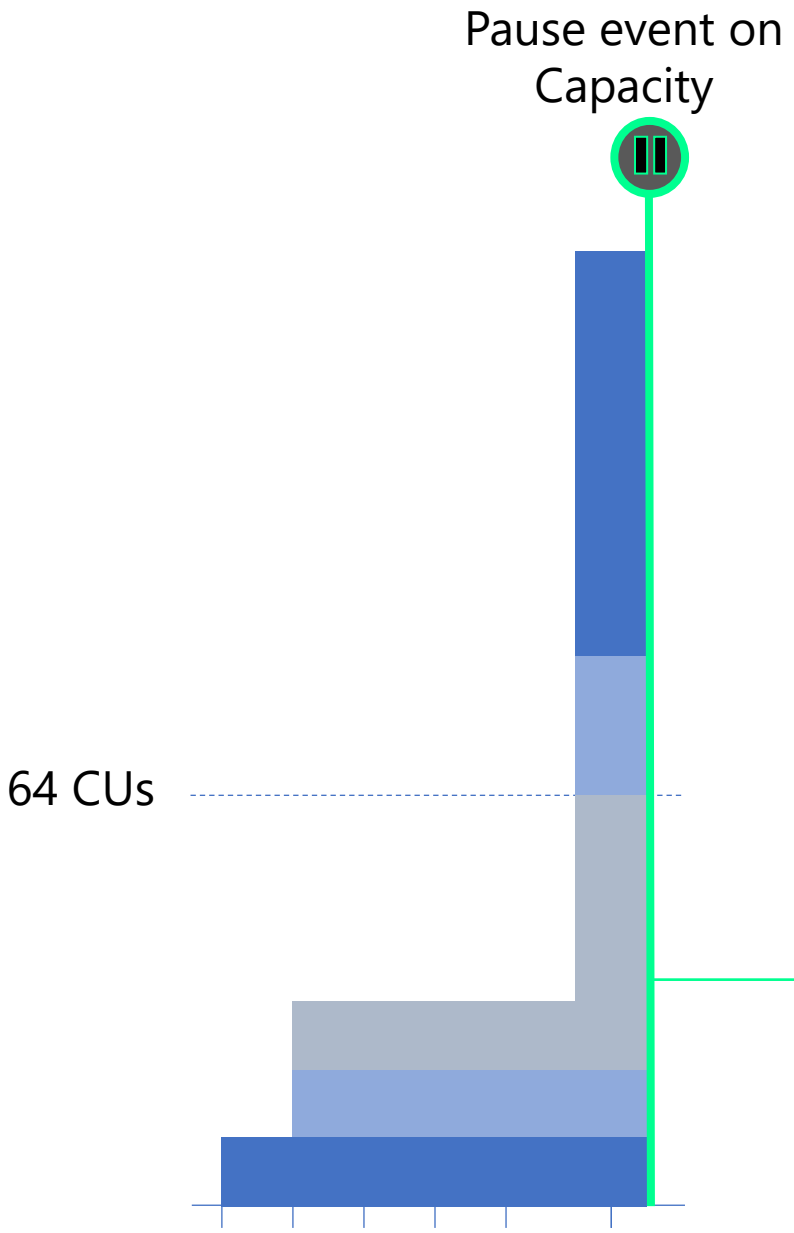


# Smoothing and Paused Capacities

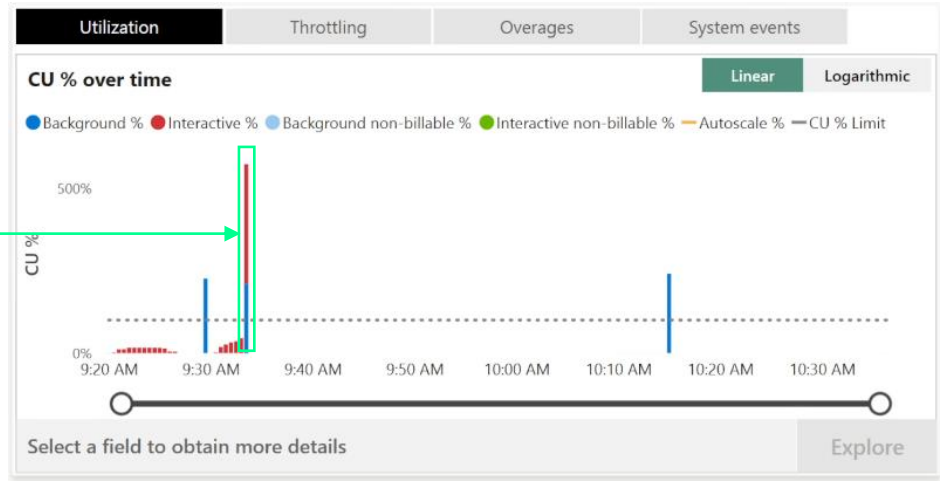
- When a capacity is paused…



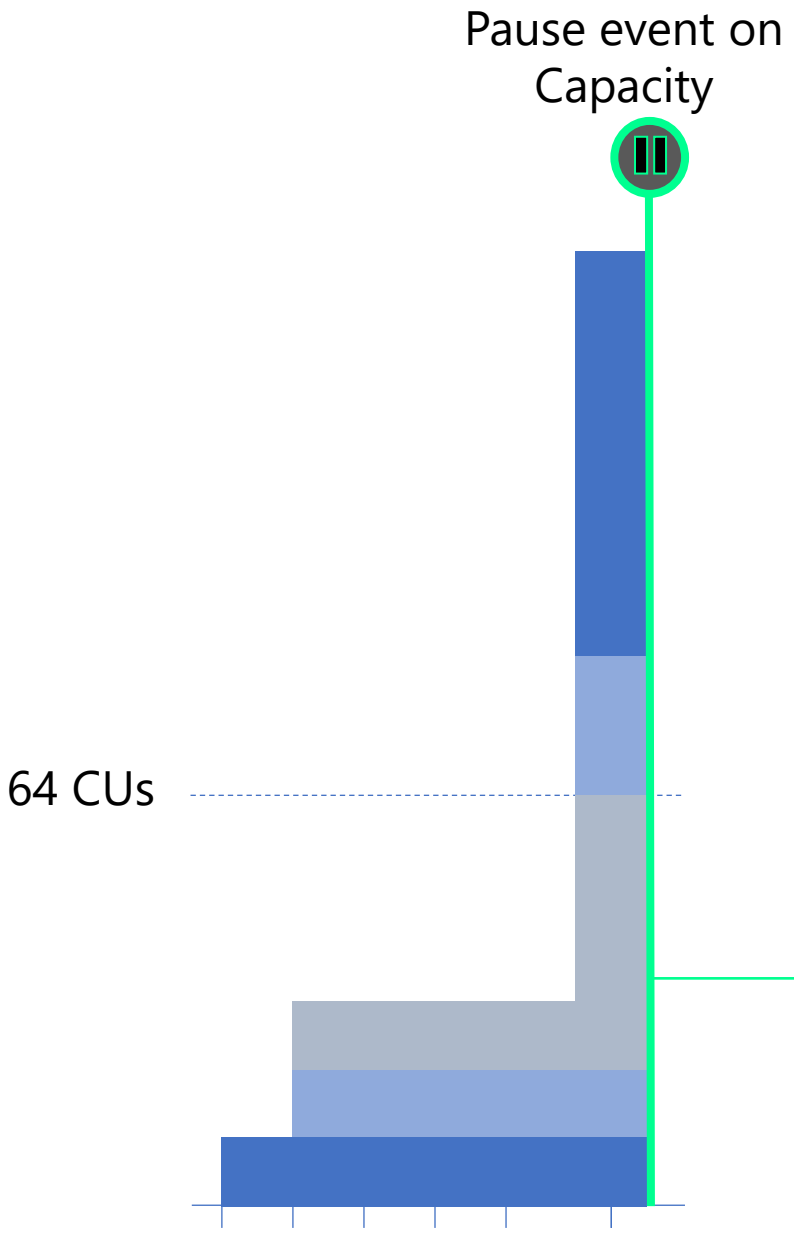
# Smoothing and Paused Capacities



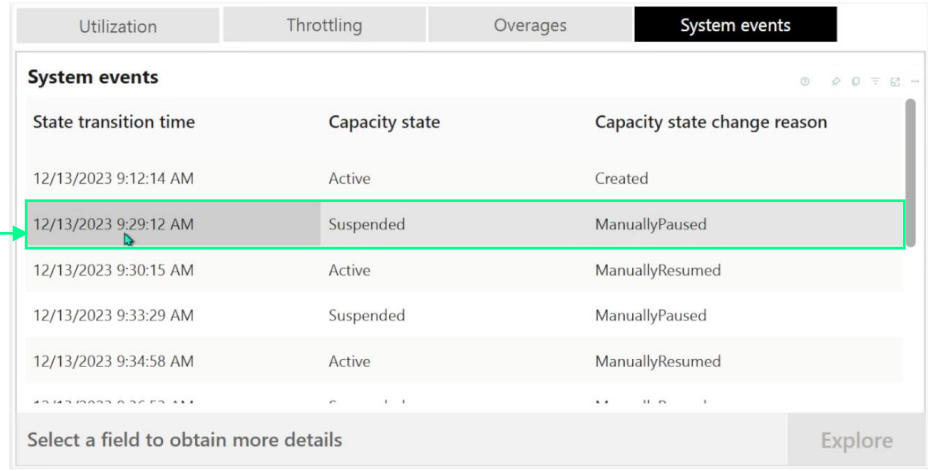

- When a capacity is paused…
- Usage that was smoothed into the future will be “reconciled” and charged against the capacity at the timestamp the capacity was paused
- Reconciled usage will show up as a spike in capacity metrics



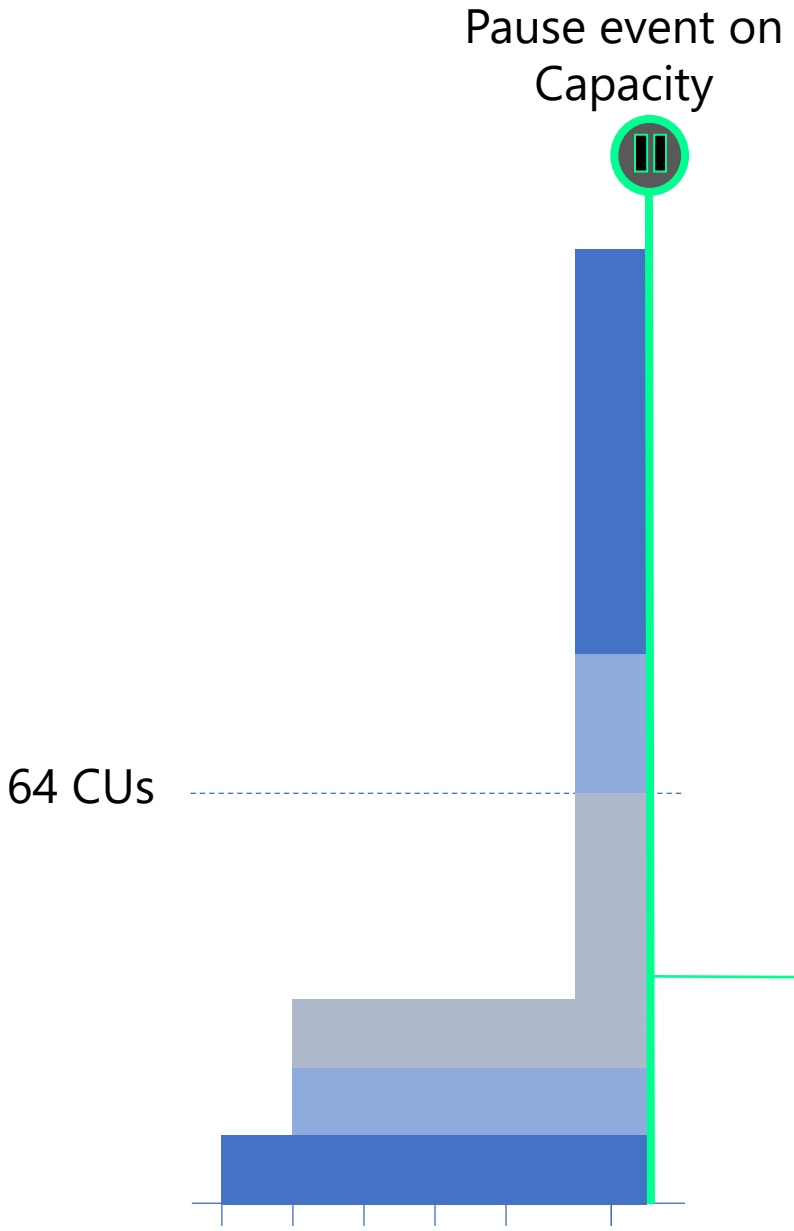
# Smoothing and Paused Capacities



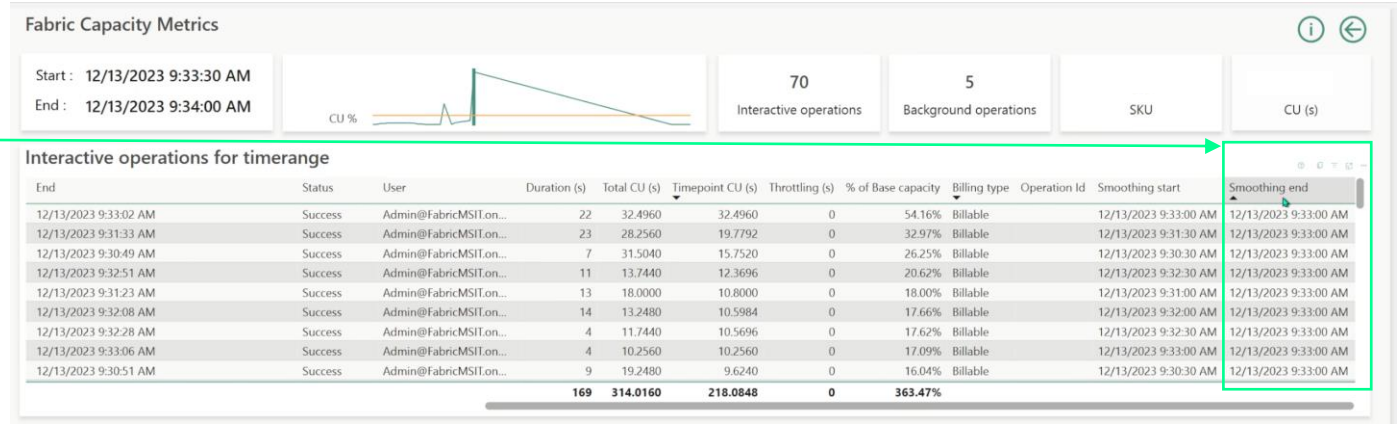

- When a capacity is paused…
- Usage that was smoothed into the future will be “reconciled” and charged against the capacity at the timestamp the capacity was paused
- Pause events can be viewed in the new System events tab



# Smoothing and Paused Capacities




- When a capacity is paused…
- Usage that was smoothed into the future will be “reconciled” and charged against the capacity at the timestamp the capacity was paused
- Pause events timestamp is shown in the smoothing end field in timepoint drill views




| End | Status | User | Duration (s) | Total CU (s) | Timepoint CU (s) | Throttling (s) | % of Base capacity | Billing type | Operation Id | Smoothing start | Smoothing end |
|---|---|---|---|---|---|---|---|---|---|---|---|
| 12/13/2023 9:33:02 AM | Success | Admin@FabricMSIT.on... | 22 | 32.4960 | 32.4960 | 0 | 54.16% | Billable | | 12/13/2023 9:33:00 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:31:33 AM | Success | Admin@FabricMSIT.on... | 23 | 28.2560 | 19.7792 | 0 | 32.97% | Billable | | 12/13/2023 9:31:30 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:30:49 AM | Success | Admin@FabricMSIT.on... | 7 | 31.5040 | 15.7520 | 0 | 26.25% | Billable | | 12/13/2023 9:30:30 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:32:51 AM | Success | Admin@FabricMSIT.on... | 11 | 13.7440 | 12.3696 | 0 | 20.62% | Billable | | 12/13/2023 9:32:30 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:31:23 AM | Success | Admin@FabricMSIT.on... | 13 | 18.0000 | 10.8000 | 0 | 18.00% | Billable | | 12/13/2023 9:31:00 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:32:08 AM | Success | Admin@FabricMSIT.on... | 14 | 13.2480 | 10.5984 | 0 | 17.66% | Billable | | 12/13/2023 9:32:00 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:32:28 AM | Success | Admin@FabricMSIT.on... | 4 | 11.7440 | 10.5696 | 0 | 17.62% | Billable | | 12/13/2023 9:32:30 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:33:06 AM | Success | Admin@FabricMSIT.on... | 4 | 10.2560 | 10.2560 | 0 | 17.09% | Billable | | 12/13/2023 9:33:00 AM | 12/13/2023 9:33:00 AM |
| 12/13/2023 9:30:51 AM | Success | Admin@FabricMSIT.on... | 9 | 19.2480 | 9.6240 | 0 | 16.04% | Billable | | 12/13/2023 9:30:30 AM | 12/13/2023 9:33:00 AM |
| | | | **169** | **314.0160** | **218.0848** | **0** | **363.47%** | | | | |

Microsoft Fabric
COMMUNITY CONFERENCE

# Bonus: Tips and Tricks for capacity management and monitoring

*"Fabric leverages the Power BI capacity model to bring even more powerful tools to your organization"*

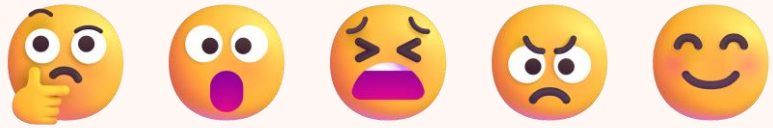
Citizen Devs

BI Specialists

Data Scientists

Data Engineers

DW/SQL Professionals

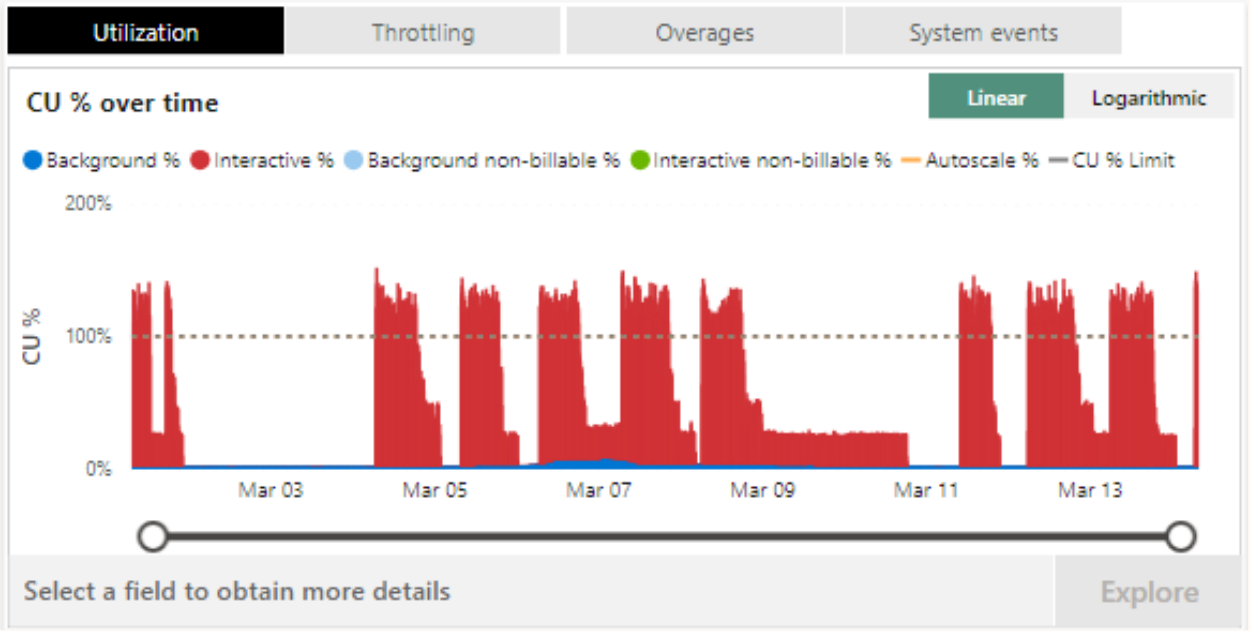


Capacity Admins

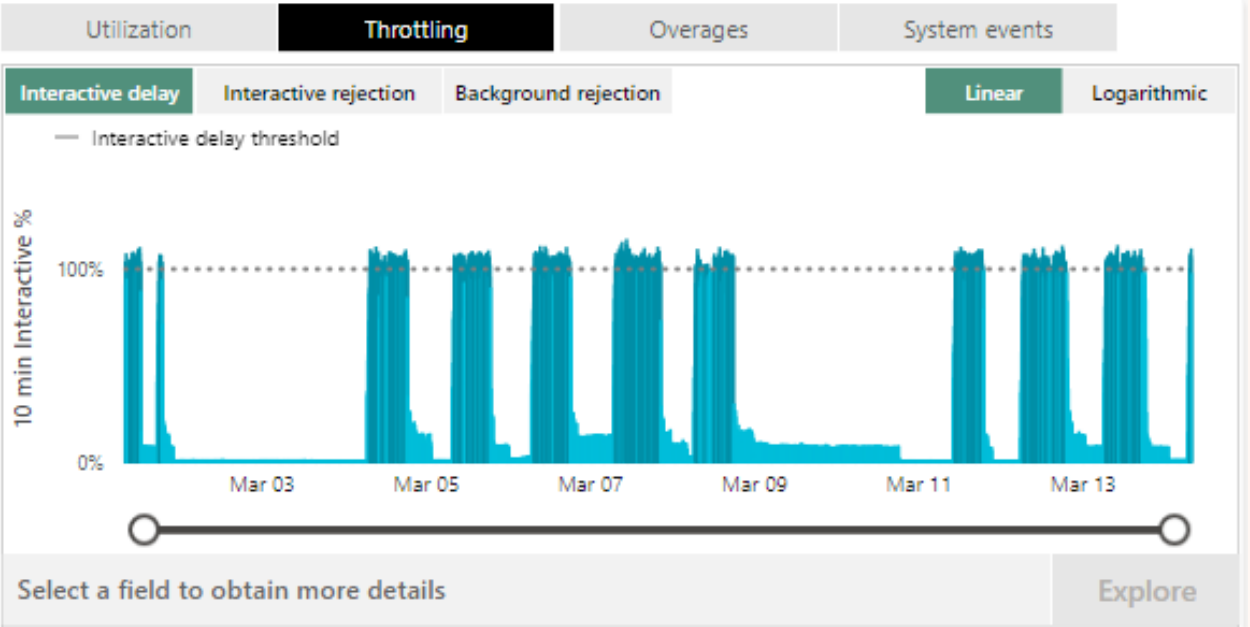
Business Leaders

Employees (Content Consumers)

# My capacity is being throttled! What can I do?



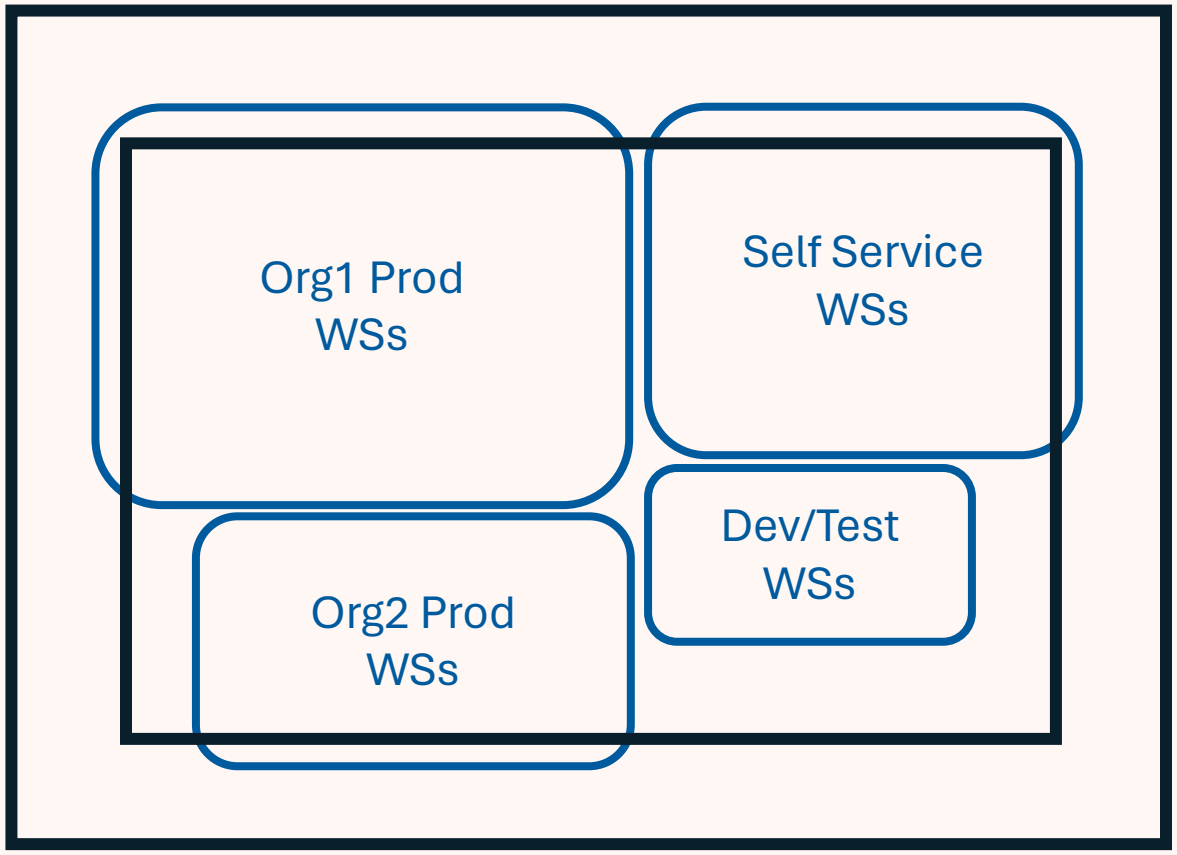

Over 100% utilization doesn't always result in throttling




No penalty until you hit 100% on one of the throttling tabs

Note: For F SKU, if throttled, you can pause/resume to pay now and clear the carry forward, but that is not a long-term solution

# When Capacity Units Run Out
# Option 1 – Scale Up




Options to add compute

- Move to a bigger P SKU or RI F SKU
- Turn on autoscale (P SKU)
- Manual/Dynamic change size (F SKU)

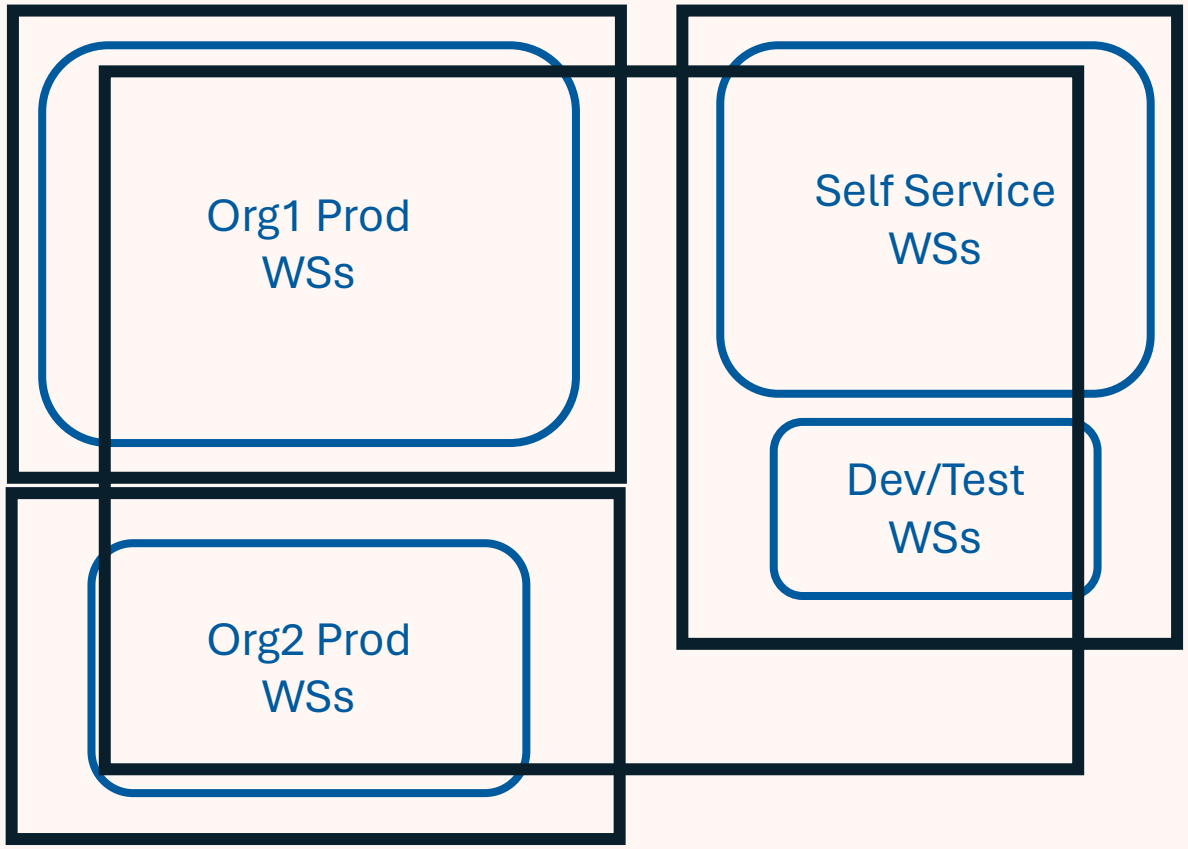
Pros

- Add CUs for all items
- Easy

Cons

- Cost
- Bad actors (items with unintentionally high CU burn) can still be a problem

# When Capacity Units Run Out
# Option 2 – Scale Out

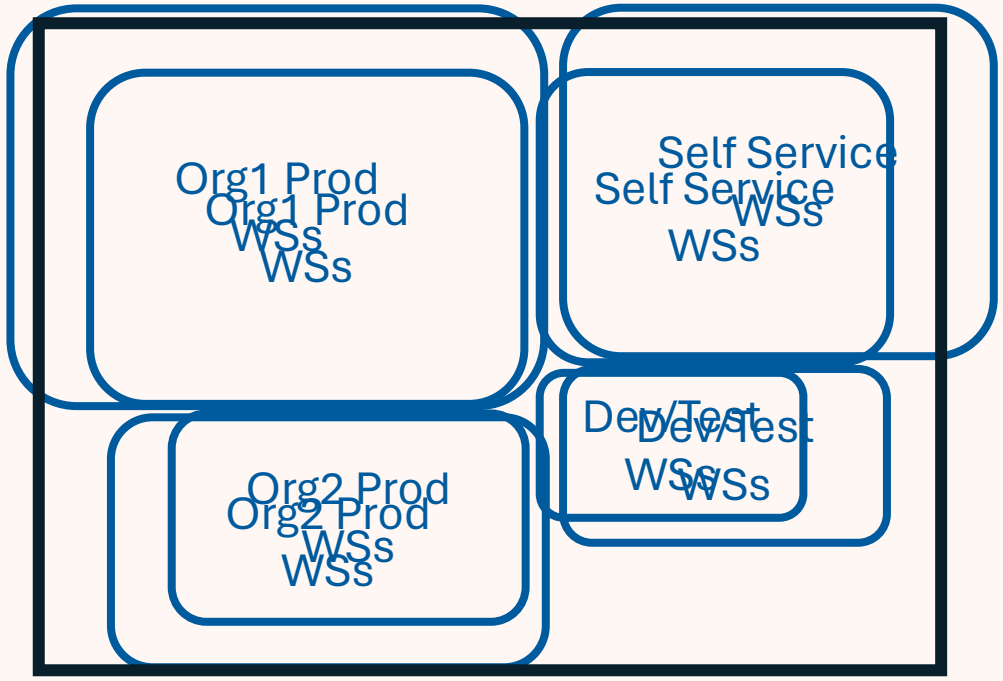



Options
- Create multiple smaller P or F SKUs based on organization, type of work, etc.

Pros
- Easy
- Provides some isolation from bad actors (items with unintentionally high CU burn)
- Flexibility in capacity settings/governance

Cons
- Cost
- High CU items have increased chance of throttling

# When Capacity Units Run Out
# Option 3 – Optimize




Approach

- Work with content creators to follow best practices and reduce CU consumption

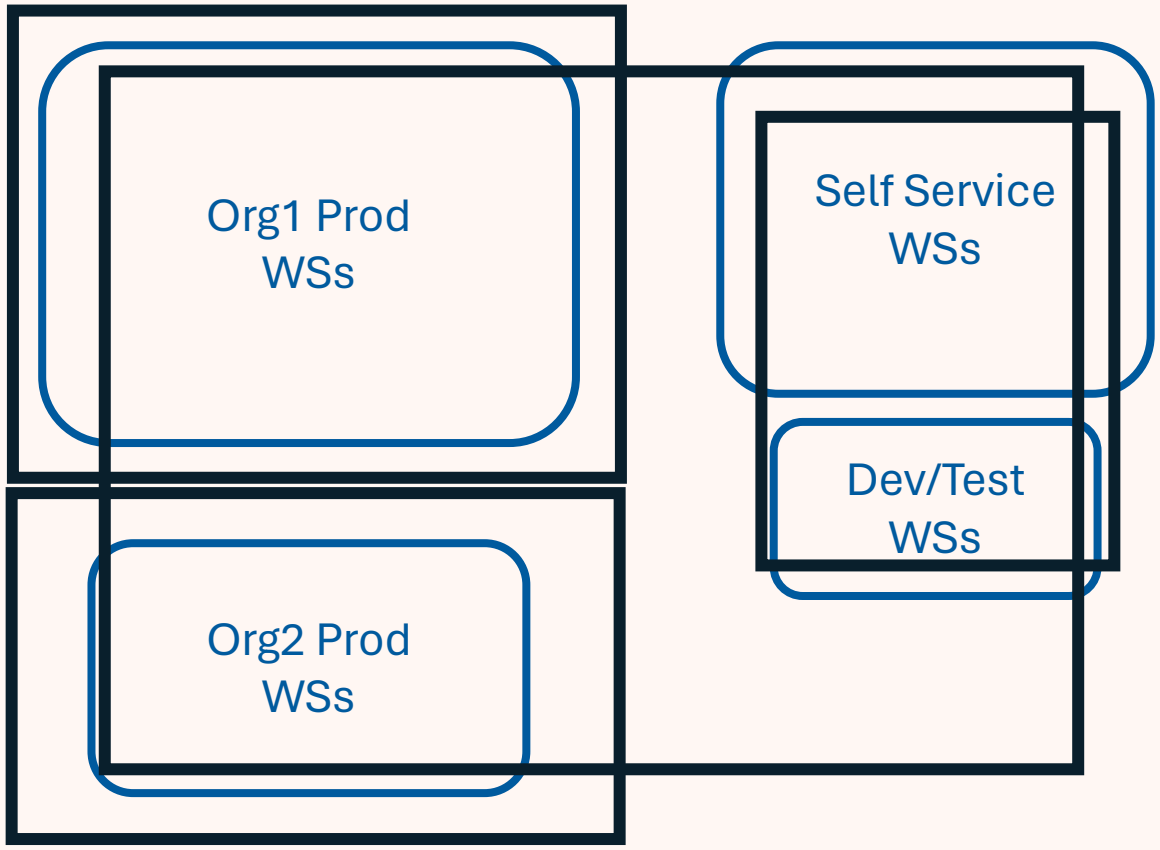
Pros

- Avoids increased cost
- Learning carries over to future content

Cons

- Can be difficult/time consuming

# When Capacity Units Run Out
# Option 4 – Isolate

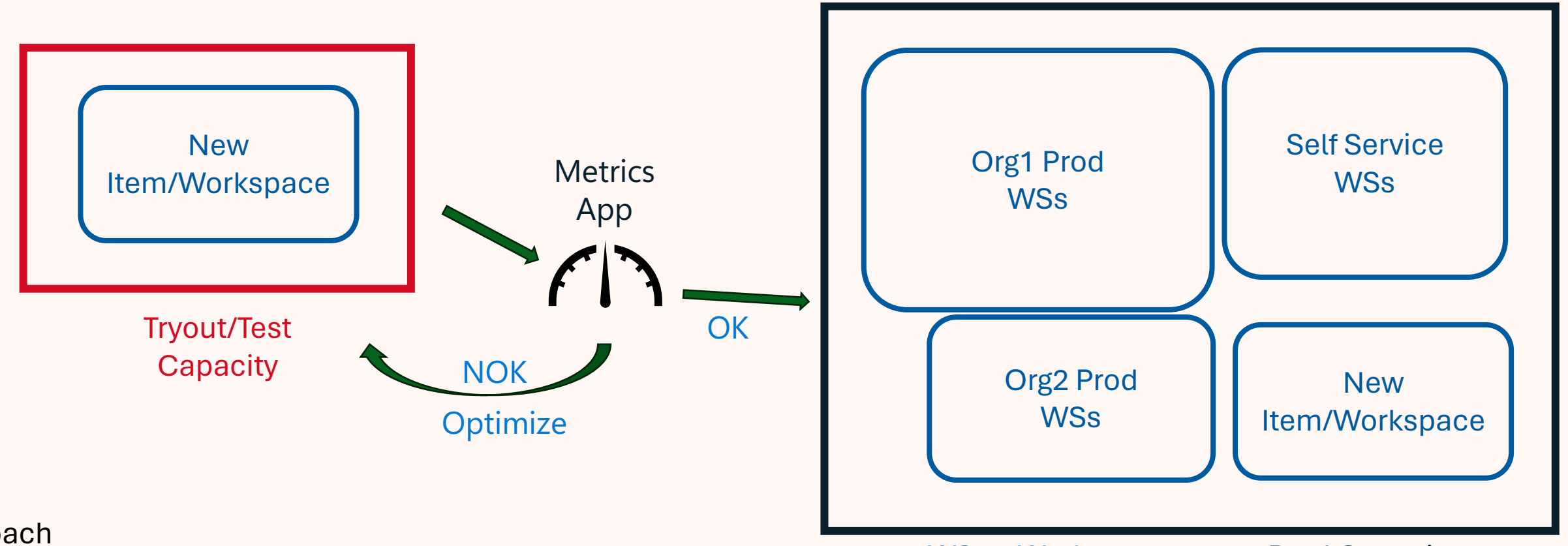



Approach
- Provide isolated capacity for key items built by experienced developers

Pros
- Easy
- Provides isolation from items built by inexperienced developers and/or rapid unplanned usage growth
- Flexibility in capacity settings/governance

Cons
- Cost
- May lead to frustration of lower priority content developers/consumers

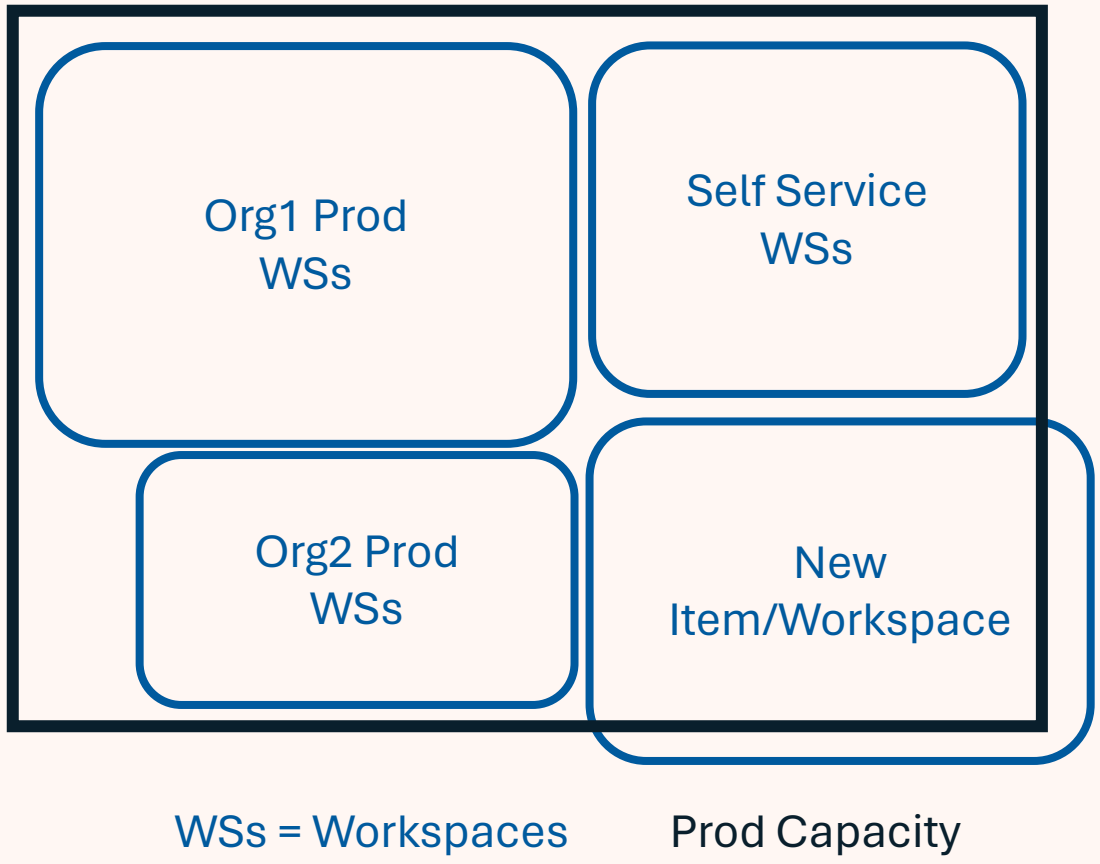
# Isolation Strategy #4a – Tryout Capacity




Approach

- Create a small F SKU capacity to “tryout” new workspaces/items
- Assess CU consumption using metrics app
- If acceptable, move to prod capacity
- If not, optimize
- Pause tryout capacity when not in use, if possible
- Note size limits for semantic model size

# Isolation Strategy #4b – Timeout Capacity

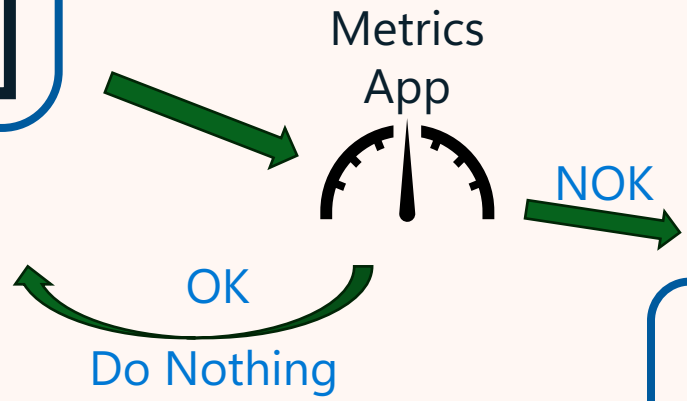


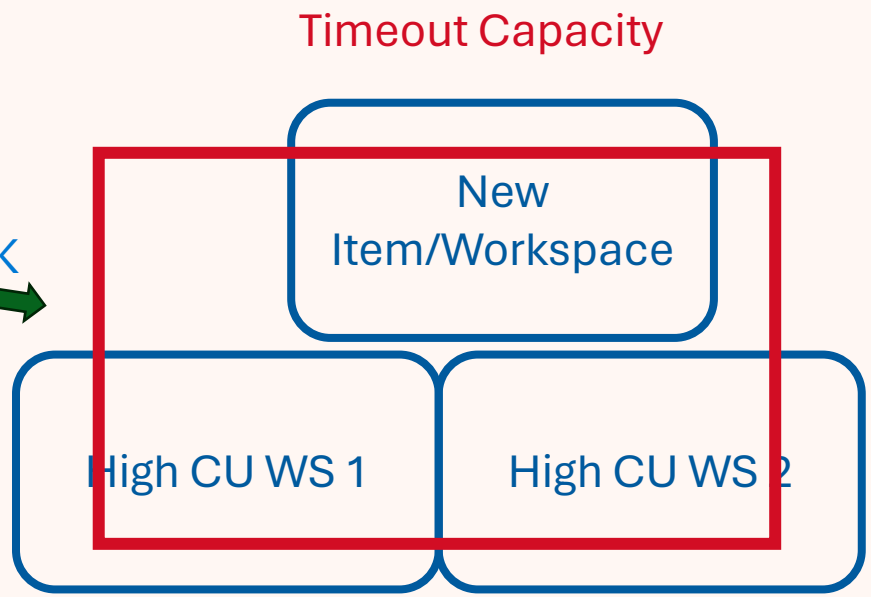

WSs = Workspaces

Prod Capacity

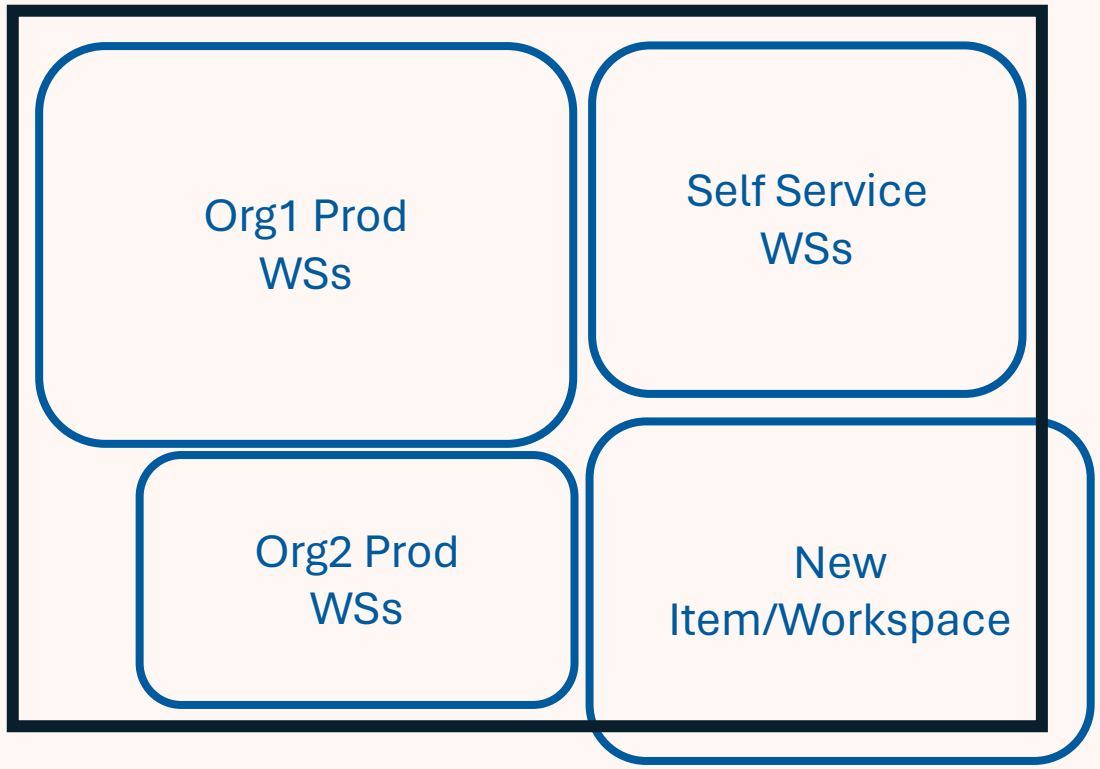
Approach

- Create a small F SKU capacity
- Assess CU consumption using metrics app
- If CU for new items/workspaces affects existing workloads (throttling), move WS to timeout capacity (Admin Portal/Capacity Settings)
- High CU items/WSs share smaller capacity (or you can pause it post move)
- Note size limits for semantic model size



# Isolation Strategy #4c – Rescue Capacity

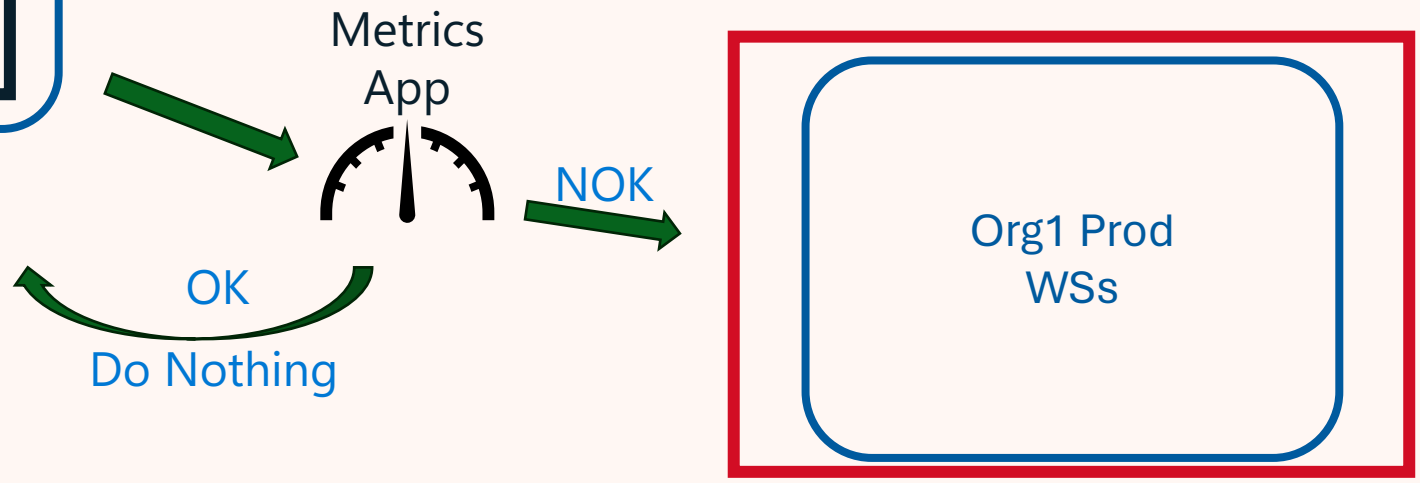



WSs = Workspaces

Prod Capacity

Approach

- Create an F SKU capacity, keep it paused
- Assess CU consumption using metrics app
- If CU for new items/workspaces affects priority workloads (throttling), resume the new capacity and move priority WS to it (Admin Portal/Capacity Settings)
- Address issues with new content, then bring it back to original capacity, and pause the new one
- Note size limits for semantic model size

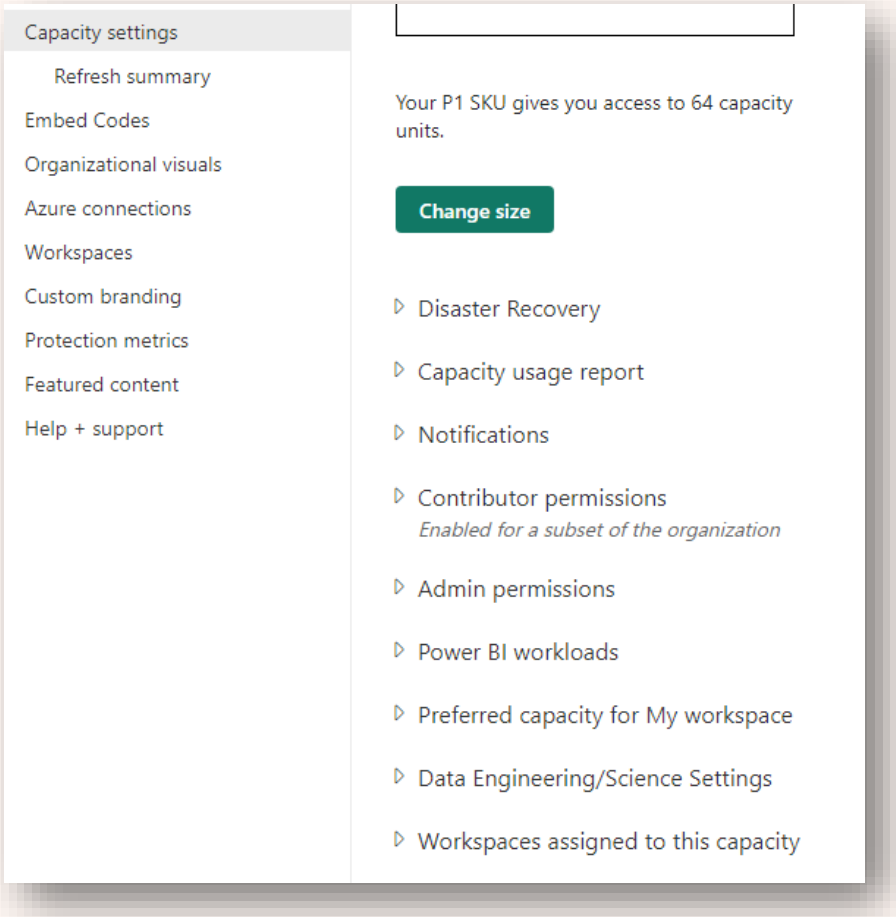


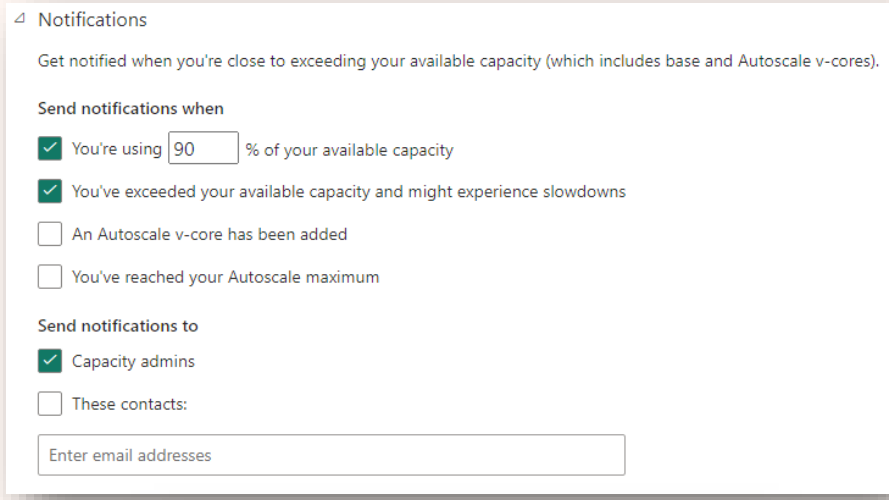
# Recommendations for Cost/CU Savings

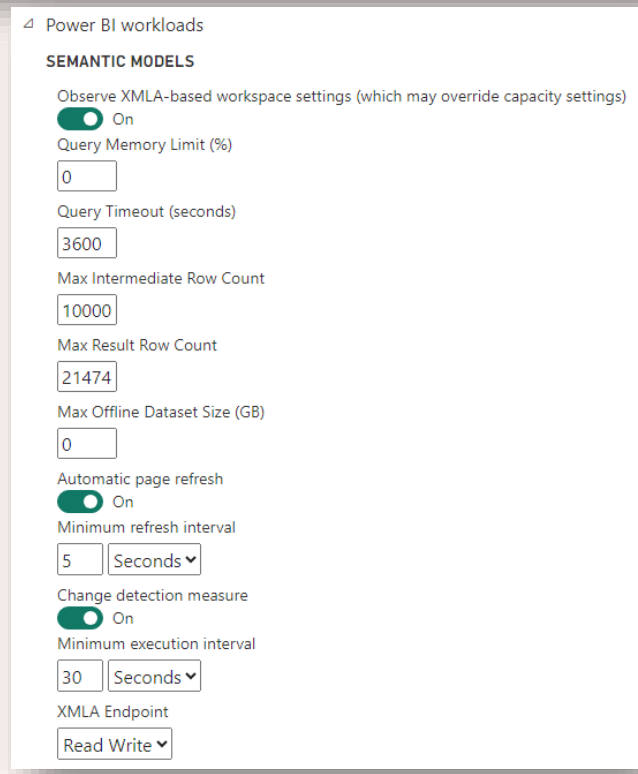
- Invest in education, knowledge/best practice sharing, COE, etc. for creators and consumers (proactive optimization)
- Avoid data/report sprawl (leverage certified/promoted models, OneLake shortcuts, etc.)
- Leverage a multi-capacity strategy (isolate, tryout, timeout, etc.)
- Right size your capacities and leverage F SKUs for pause/resume/resize, or reserved instances for discounts
  - Consider a combo of RI and PAYGO (for predictable surge activity)
- Choose the right tool for the job and stay up to date on Fabric feature releases
  - High concurrency mode for notebooks

# Leverage the capacity settings in the UI

- Notifications on CU overuse
- Power BI workloads settings (e.g., query limits, page refresh)



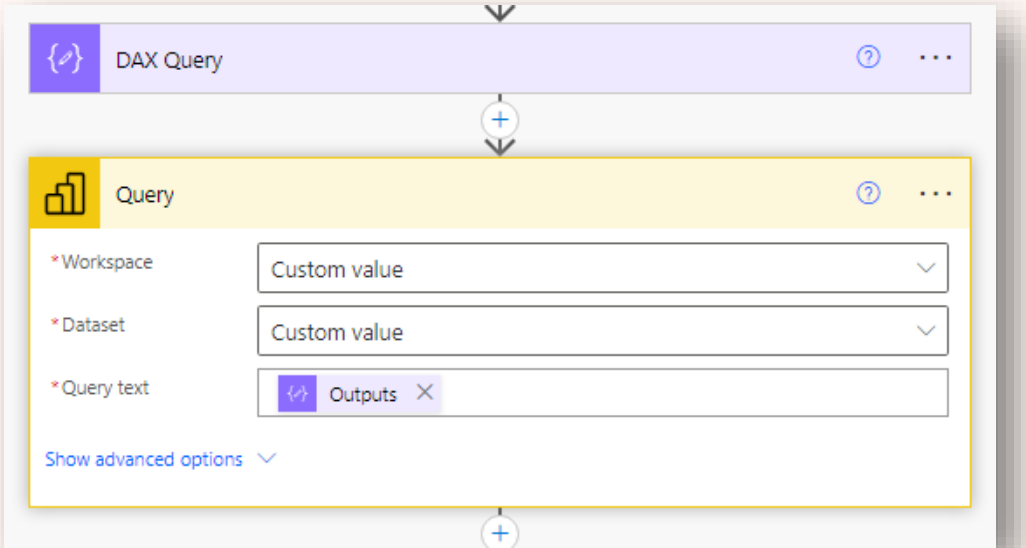



# Custom Solutions

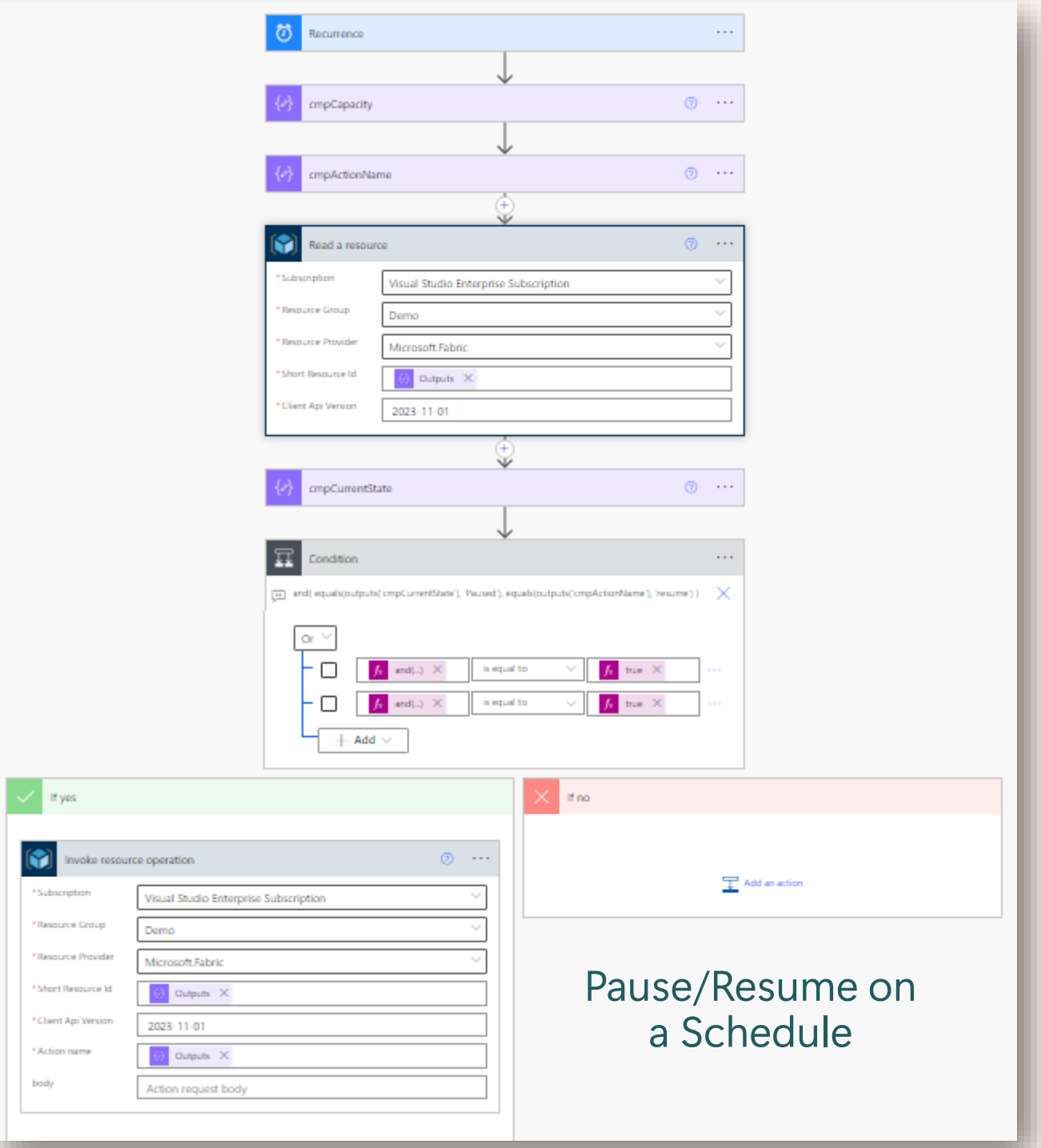
- Modify the Metrics App to meet your needs
- Build a custom report off the semantic model
- Send DAX queries to the metrics app semantic model in your own solution
  - Power Automate, Notebook (SemPy), PowerShell, etc.
  - Get throttling % values (Interactive Delay, Interactive Rejection, and/or Background Rejection)
  - Latest values and/or trends over time
  - Best for summarized data only (e.g., hour, day)

Incorporate Metrics App queries into custom solutions




Collect data from multiple capacities and store it long term

```python
# Get max date from current delta table (to avoid loading duplicate days)
try:
    df_max = spark.sql(f'''
    SELECT MAX(Date) as MaxDate
    FROM throttling;
    ''')
    maxdate = df_max.first()['MaxDate']
except:
    maxdate = datetime.today() + timedelta(days=-6)
maxdateforDAX = maxdate.strftime('%Y,%m,%d')

if maxdate.date() < (datetime.today() + timedelta(days=-1)).date():

    # Get data for each capacity, write daily csv and append delta
    for capacity in lst_capacities:
        querytext = '''\
                    DEFINE
                    MPARAMETER 'CapacityID' = "{capID}"
                    VAR yesterday =
                        FILTER(ALL('Dates'[Date] ), 'Dates'[Date] < TODAY() && 'Dates'[Date] > DATE({MD}) )

                    EVALUATE
                    SUMMARIZECOLUMNS(
                        'Dates'[Date],
                        'TimePoints'[Start of Hour],
                        yesterday,
                        "IntDelay", ROUND( 'All Measures'[Dynamic InteractiveDelay %] * 100, 2 ),
                        "IntReject", ROUND( 'All Measures'[Dynamic InteractiveRejection %] * 100, 2 ),
                        "BackReject", ROUND( 'All Measures'[Dynamic BackgroundRejection %] * 100, 2 )
                    )
                    '''.format(capID=capacity, MD=maxdateforDAX)
        df_throttling = fabric.evaluate_dax(workspace=MetricsWS, dataset=MetricsModel, dax_string=querytext)
        if len(df_throttling) >= 1:
            df_throttling.columns = df_throttling.columns.str.replace(r'(.*\[)|(\].*)', '', regex=True)
            df_throttling.columns = df_throttling.columns.str.replace(' ', '_')
            df_throttling['capacityId'] = capacity
            filename = capacity + '_throttling_' + (datetime.today()).strftime('%Y%m%d') + '.csv'
            df_throttling.to_csv("/lakehouse/default/Files/ThrottlingData/" + filename)
            spk_throttle = spark.createDataFrame(df_throttling)
            spk_throttle.write.mode("append").format("delta").option("overwriteSchema", "true").saveAsTable('Throttling')
```




Pause/Resume on a Schedule

# Automate With F SKUs

- Pause/resume on a schedule
  - Automate with Power Automate, Logic Apps, or a Notebook
- Resize at peak/slow times
  - Mix with Reserved Instance (PAYGO when at increased size)
  - Query the metrics app and respond to actual demand (DIY autoscale)

# DIY Autoscale – Fabric Notebook

(Bret Myers)

## Set SKU Ranges and Values

```
# Parameters to be passed in from pipeline.
minSku = 'F2' # min sku size we can scale down to
maxSku = 'F128' # max sku size we can scale up to
utilizationTolerance = 90 # Percentage of CU used to st
capacityName = 'fabricbamdemo' #capacity name to be mon
subscriptionId = '[redacted]'
metricsAppWorkspaceName = 'WS_FabricCapacityMetrics' #
metricsAppModelName = 'Fabric Capacity Metrics' # name
alertEmail = '' # email address to send alert that we s
```

## Get credentials

```
tenantId = mssparkutils.credentials.getSecret('keyVaultEndpoint', 'secretName_tenantId')
clientId = mssparkutils.credentials.getSecret('keyVaultEndpoint', 'secretName_clientId')
secret = mssparkutils.credentials.getSecret('keyVaultEndpoint', 'secretName_clientSecret')

api_pbi = 'https://analysis.windows.net/powerbi/api/.default'
api_azuremgmt = 'https://management.core.windows.net/.default'
```

## Not all code shown

FabricTools/CapacityAutoScale at main · bretamyers/FabricTools · GitHub

## Query metrics app model

```
from azure.identity import ClientSecretCredential
import requests, json, math
from pyspark.sql.functions import explode

auth = ClientSecretCredential(tenant_id=tenantId, client_id=clientId, client_secret=secret)
access_token = auth.get_token(api_pbi)

header = {'Authorization': f'Bearer {access_token.token}', 'Content-type': 'application/json'}

body = {
  "queries": [
    {
      "query": f"""
        DEFINE
          MPARAMETER 'CapacityID' = "{capacityId}"

          VAR __DS0FilterTable =
            FILTER(
              KEEPFILTERS(VALUES('TimePoints'[TimePoint])),
              'TimePoints'[TimePoint] >= NOW() - 1
            )

          VAR __DS0FilterTable2 = TREATAS({{"{capacityId}"}}, 'Capacities'[capacityId])

          VAR __DS0Core =
            SELECTCOLUMNS(
              KEEPFILTERS(
                FILTER(
                  KEEPFILTERS(
                    SUMMARIZECOLUMNS(
                      'Capacities'[capacityId],
                      'Items'[Billable type]
```

## Change SKU Size

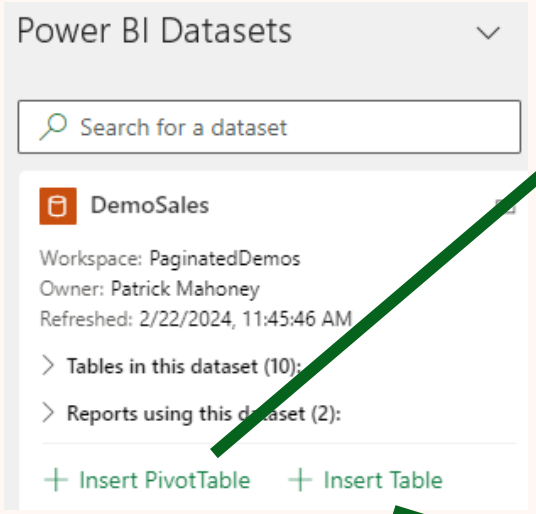
```
    url = f'https://management.azure.com/subscriptions/{subscriptionId}/resourceGroups/{res

    body = {
                "sku": {
                    "name": f"{scaleSku}",
                    "tier": "Fabric"
                }
            }

    response = requests.patch(url, headers=header, data=json.dumps(body))
```

# Most Common Capacity Issues (Power BI)

| Bad Practice | Recommendations/Typical Resolution |
|---|---|
| Model issues (M:M, bi-di, snowflake, etc.) and/or inefficient DAX | Follow best practices (e.g., BPA), star schema |
| Too many visuals | Multi card, small multiples, Deneb, PowerPoint background, etc. |
| Big single visual (i.e., matrix with lots of rows, columns, and/or measures) | Improve report design (e.g., drillthrough, apply all Slicers, report page tooltip), field parameters, calc group guardrails, etc. |
| Complex RLS | Remodel to enable simple filter like Table[Email] = USERPRINCIPALNAME() |
| Very high concurrency | Optimize reports, DAX, etc. (big multiplier)<br>Consider QSO, data subsets |
| Direct Query | Switch to import or Direct Lake, if possible. Aggregations, hybrid tables, etc. |
| Analyze in Excel | Automate downstream analytics with a Power BI report instead, subscriptions, DAX connected table, slicers/measures first, etc. |
| Excessive refresh | Don’t “break the fold”, incremental refresh, reduce frequency, optimize M code |

# Save Those CUs – Getting Data Into Excel

## Analyze in Excel




## Connected Table

## ❌ Rows, Measures, Filter

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 11:49:30 | MDX | 2,328ms | Power BI... | DemoSales | SELECT {[Measures].[To |
| 11:49:26 | MDX | 0ms | Power BI... | DemoSales | SELECT {AddCalculated |
| 11:49:23 | MDX | 0ms | Power BI... | DemoSales | SELECT {AddCalculated |
| 11:49:17 | MDX | 1,875ms | Power BI... | DemoSales | SELECT {[Measures].[To |
| 11:49:03 | MDX | 4,469ms | Power BI... | DemoSales | SELECT {[Measures].[To |
| 11:48:54 | MDX | 3,938ms | Power BI... | DemoSales | SELECT {[Measures].[To |

## ✅ Filter, measures, rows

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 10:06:13 | MDX | 1,625ms | Power BI... | DemoSales | SELECT {[Measu |
| 10:06:03 | MDX | 781ms | Power BI... | DemoSales | SELECT {[Measu |
| 10:05:49 | MDX | 109ms | Power BI... | DemoSales | SELECT {[Measu |
| 10:05:46 | MDX | 312ms | Power BI... | DemoSales | SELECT {[Measu |
| 10:05:43 | MDX | 234ms | Power BI... | DemoSales | SELECT FROM [M |
| 10:05:14 | MDX | 0ms | Power BI... | DemoSales | SELECT {AddCal |

## Refresh (same for both)

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 11:50:30 | MDX | 2,234ms | Power BI... | DemoSales | SELECT {[Measures].[To |

## ❌ Rows, Measure, Filter

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 01:28:50 | DAX | 31ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:41 | DAX | 1,516ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:40 | DAX | 16ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:34 | DAX | 156ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:33 | DAX | 16ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:31 | DAX | 0ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:30 | DAX | 141ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:15 | DAX | 2,047ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:11 | DAX | 1,797ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:28:08 | DAX | 594ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:27:56 | DAX | 281ms | Power BI... | DemoSales | DEFINE VAR __D |
| 01:27:50 | DAX | 16ms | Power BI... | DemoSales | DEFINE VAR __D |

## ✅ Filter, measures, rows

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 09:14:20 | DAX | 16ms | Power BI... | DemoSales | DEFINE VAR __DS0Filte |
| 09:14:07 | DAX | 1,000ms | Power BI... | DemoSales | DEFINE VAR __DS0Filte |
| 09:14:02 | DAX | 1,188ms | Power BI... | DemoSales | DEFINE VAR __DS0Filte |
| 09:13:59 | DAX | 594ms | Power BI... | DemoSales | DEFINE VAR __DS0Filte |
| 09:13:51 | DAX | 531ms | Power BI... | DemoSales | DEFINE VAR __DS0Filte |
| 09:13:50 | DAX | 0ms | Power BI... | DemoSales | DEFINE VAR DS0Cor |

## Refresh (same for both)

| StartTime | Type | Duration | User | Database | Query |
|---|---|---|---|---|---|
| 11:54:49 | DAX | 1,969ms | Power BI... | DemoSales | DEFINE VAR __DS0FilterTable = TREATA |

Key Takeaways

- How you build it matters
  - Filters & measures first!
- This shows durations but it's CU that matters (test your use cases/models)
- Opt for DAX Connected Tables
  - Create pivot table from that, if needed


I PITY THE FOOL
THAT DOESN'T
GIVE SESSION FEEDBACK
imgflip.com

# Slides

https://github.com/BenniDeJagere/Presentations/{Year}/{YYYYMMDD}_{Event}

# Thank you