

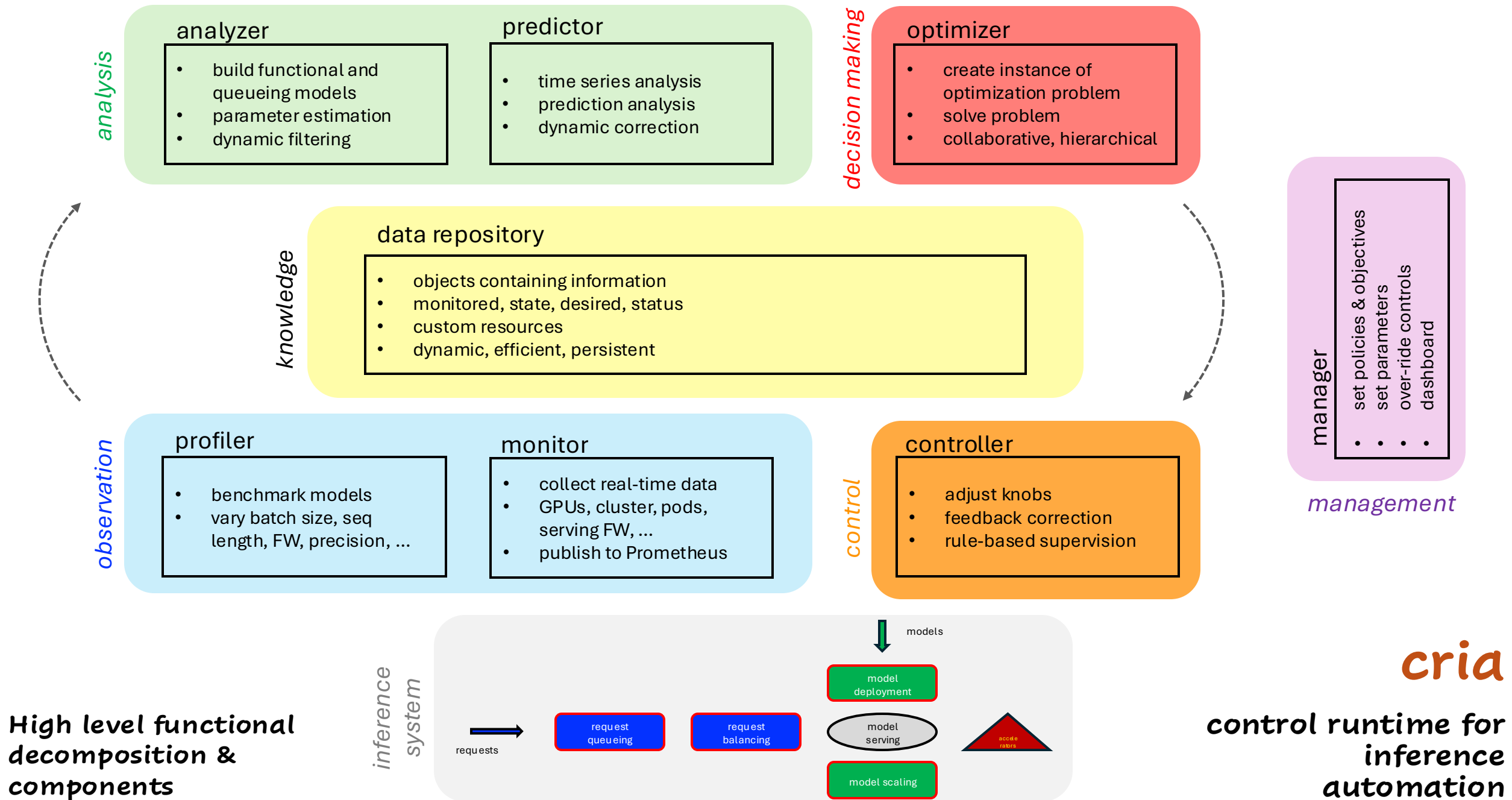
inferno
inference platform,
naturally optimized

Inferencing operational optimization

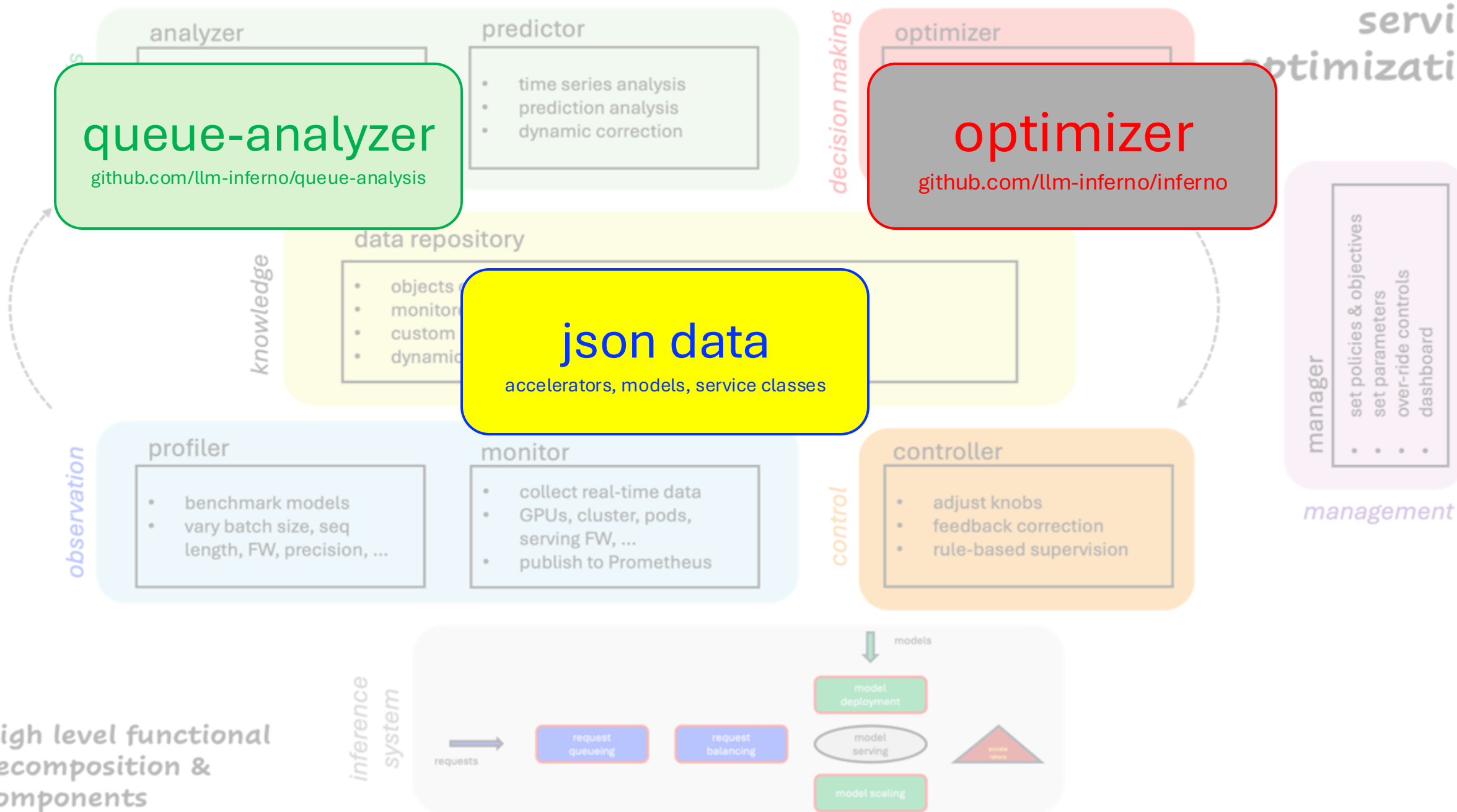
Model-based, dynamic global
optimization

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IBM Research

16 August 2024



Inference serving optimization



Inference
serving
optimization

traffic-
predictor

queue-analyzer
github.com/llm-inferno/queue-analysis

optimizer
github.com/llm-inferno/inferno

json data
accelerators, models, service classes

model-
adjuster

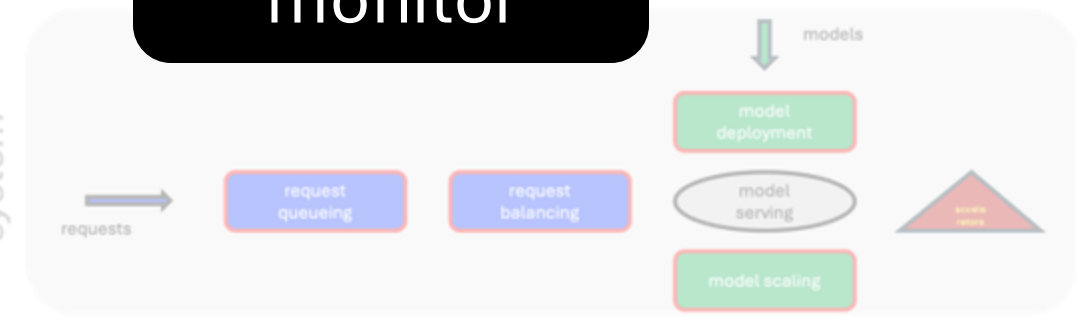
traffic-
monitor

deploy-
orchestrator

manager
• set policies & objectives
• set parameters
• over-ride controls
• dashboard
management

High level functional
decomposition &
components

inference
system



Problem scope

- memory demand
- performance characteristics (delay, throughput)

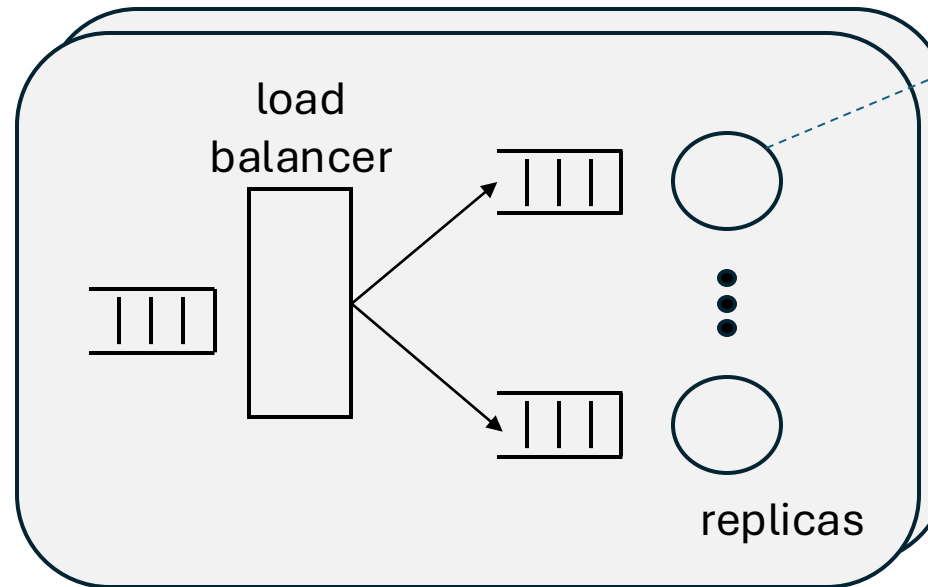
- cost
- memory size
- memory BW

models

GPU types

requests

- classes of service
- arrival patterns
- service patterns



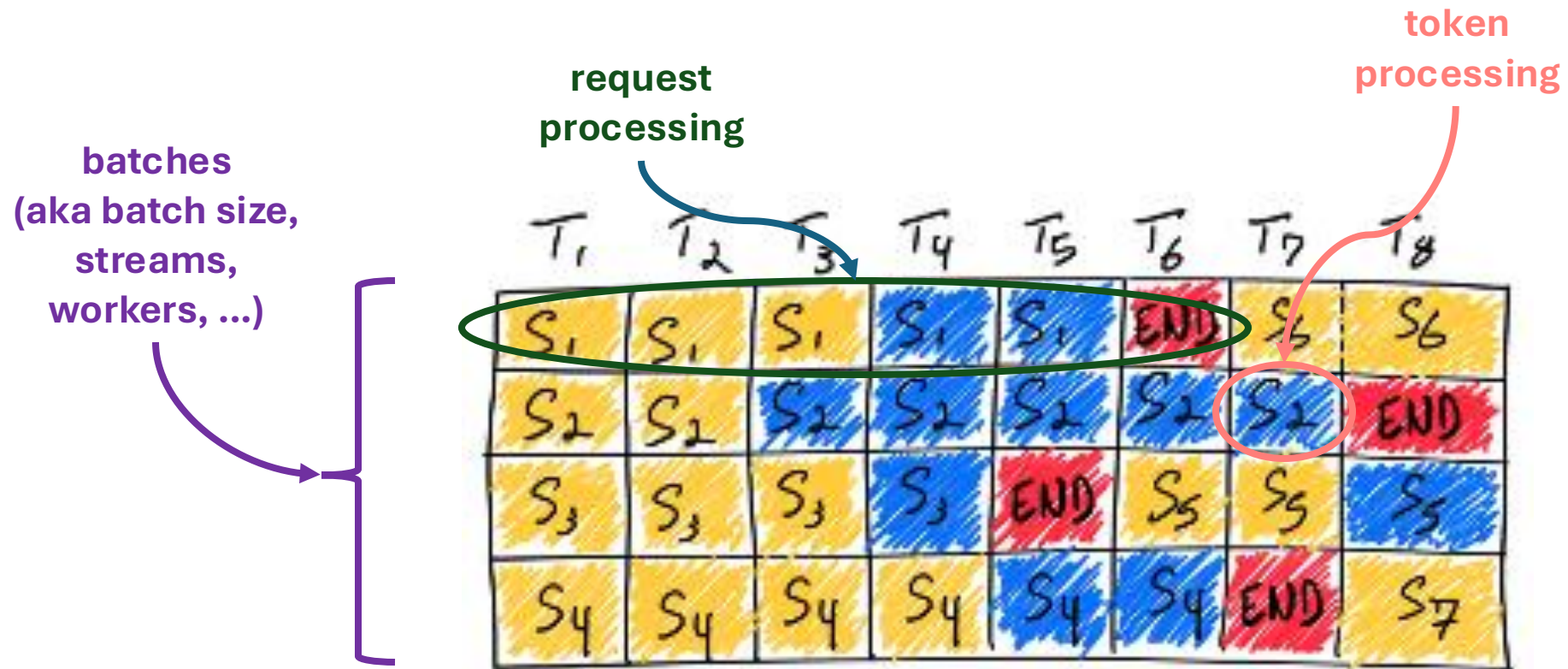
objectives

- min cost
- satisfy SLOs
- fairness, slack, power

decision variables

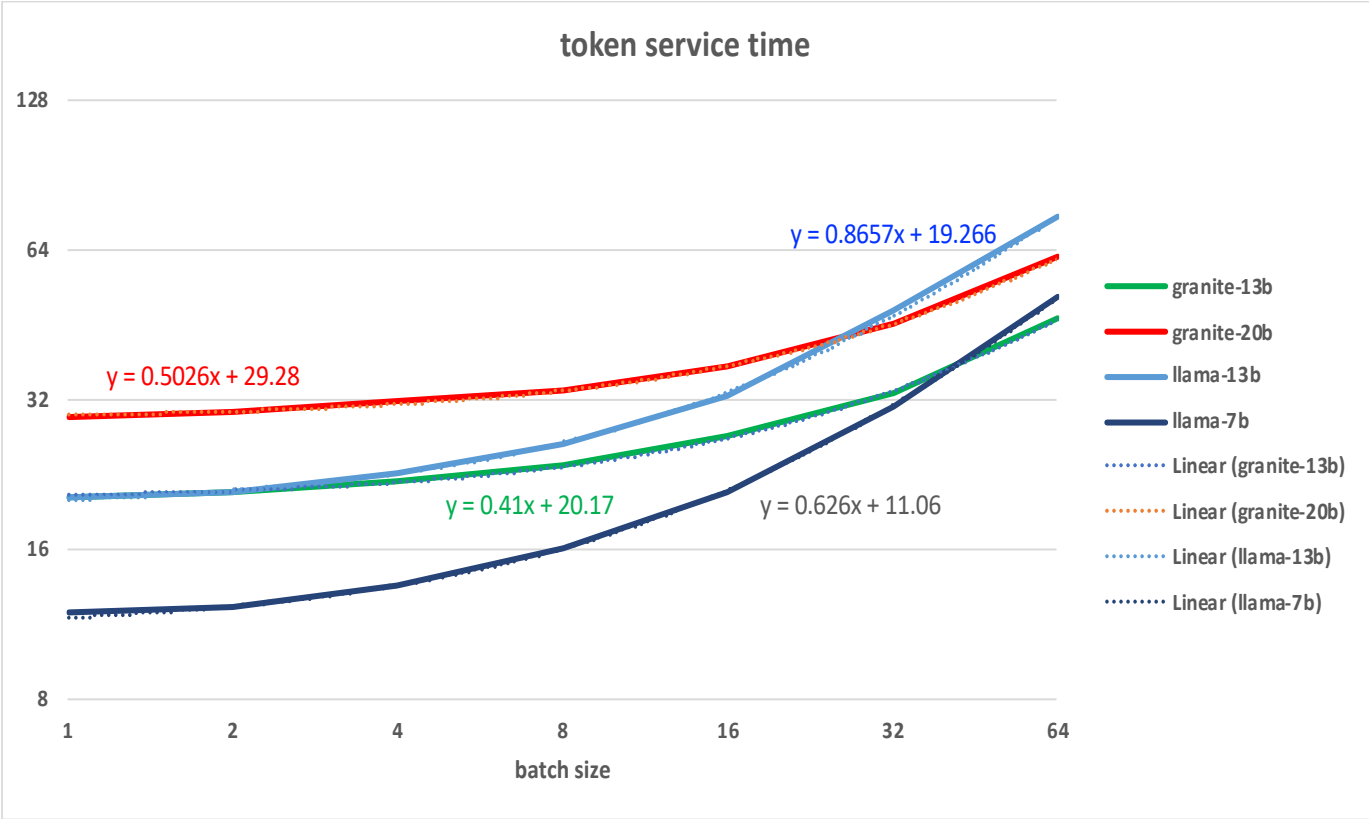
- GPU type
- number of replicas
- batch size

Request batching



N batch size
 $\tau(N)$ token service (processing) time (*ITL*)
 K number of tokens per request

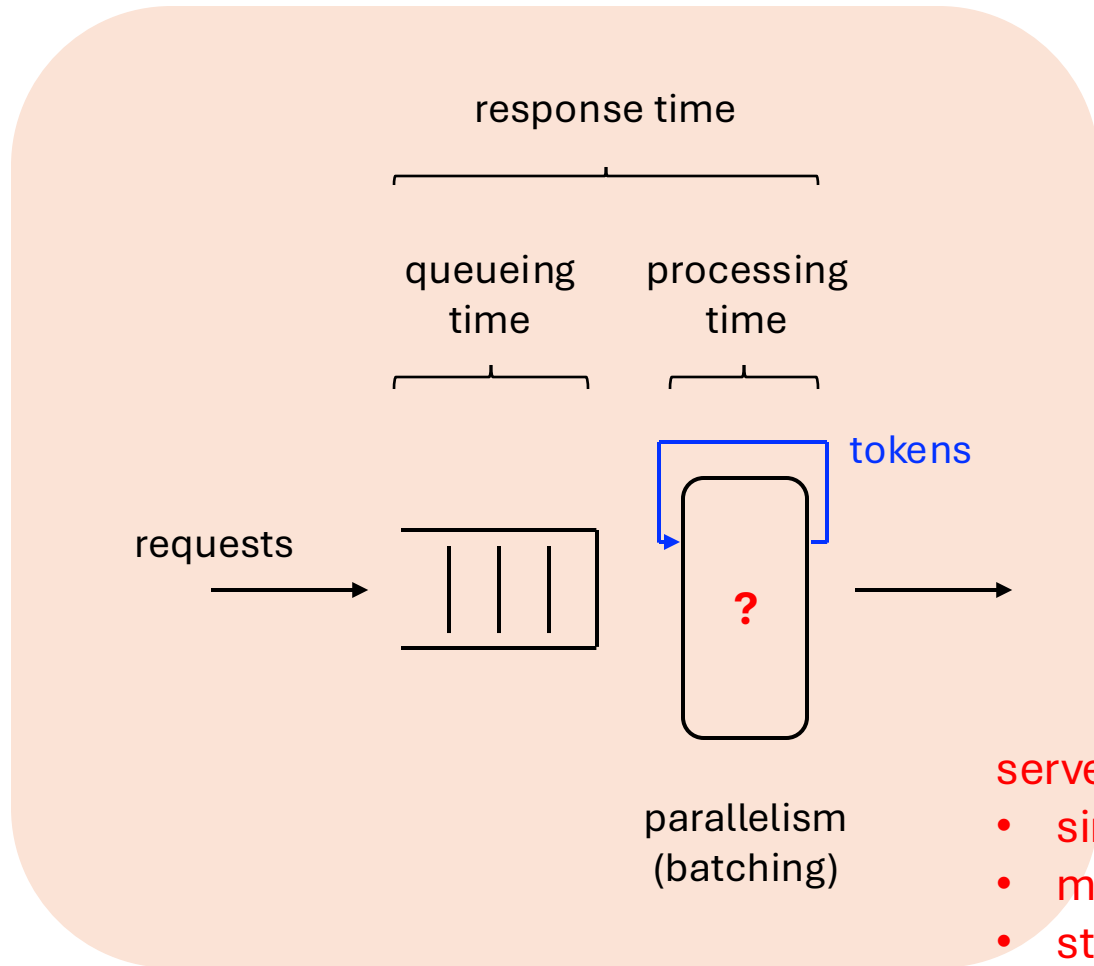
Fitting token service time



precision	fp16
accelerator	A100
tokens	1024
back	vLLM
accCount	1

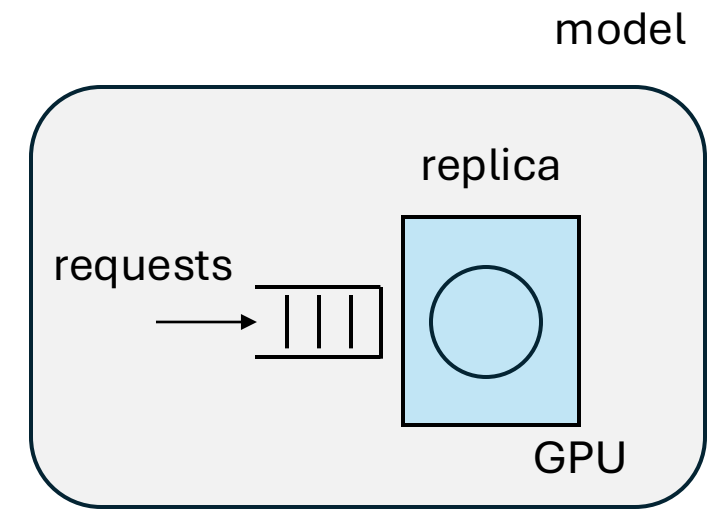
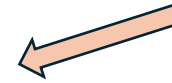
$$\tau(N) \approx \alpha + \beta N$$

Queueing model of a model serving instance (replica)



server structure?

- single server
- multiple server
- state-dependent server
- queueing network



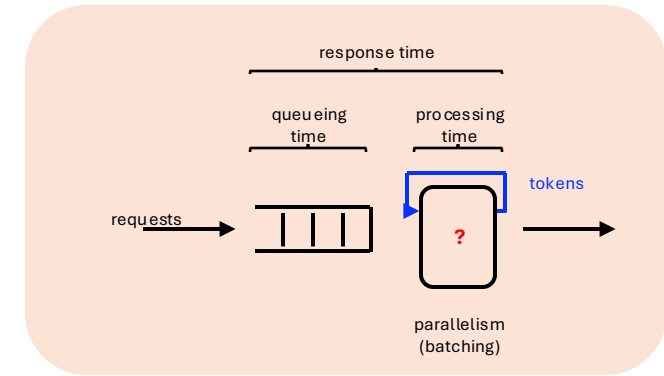
avg request processing time =
avg num tokens per request *
avg token service time

load = request rate * avg request processing time

avg token delay = avg request response time /
avg num tokens

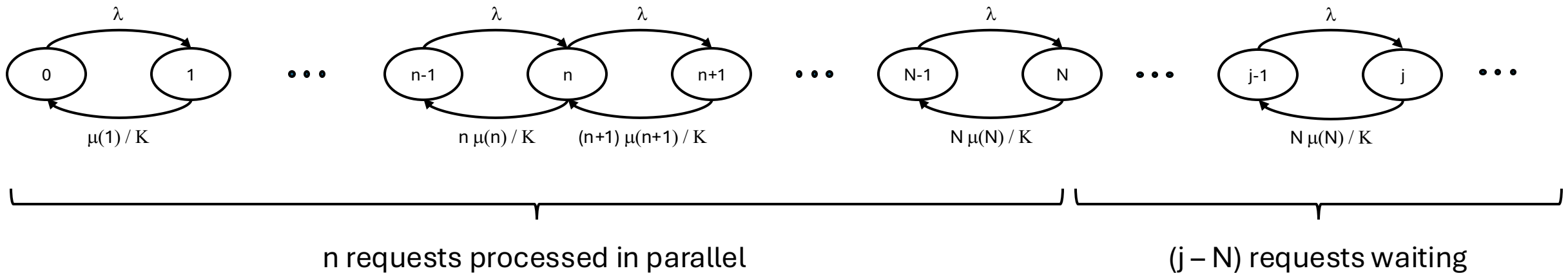
modeling batching

- Markovian assumptions
- state: number of requests in process
- λ request arrival rate
- K average number of tokens per request
- $\tau(n)$ average token service time given n batches
- N batch size (maximum)



$$\mu(n) = \frac{1}{\tau(n)}$$

$$\tau(n) \approx \alpha + \beta n$$



rate limiting

requestRate	36.613 /min	p(0)	3.0013E-09	avgNumSystem	33.77
lambda	0.00061 /msec	maxN	200	avgNumServer	33.16
K	1024	pFull	1.0758E-11	avgNumQueue	0.61
N	48	effectPut	36.61299 req/m	utilization	0.691
alpha	19		0.00061 /msec	avgRespTime	55,349
beta	1	P(N)	0.92202	avgSrvTime	54,349
lambda*K	0.62486163			avgWaitTime	1,000
				avgTokenServTime	53.1

Find

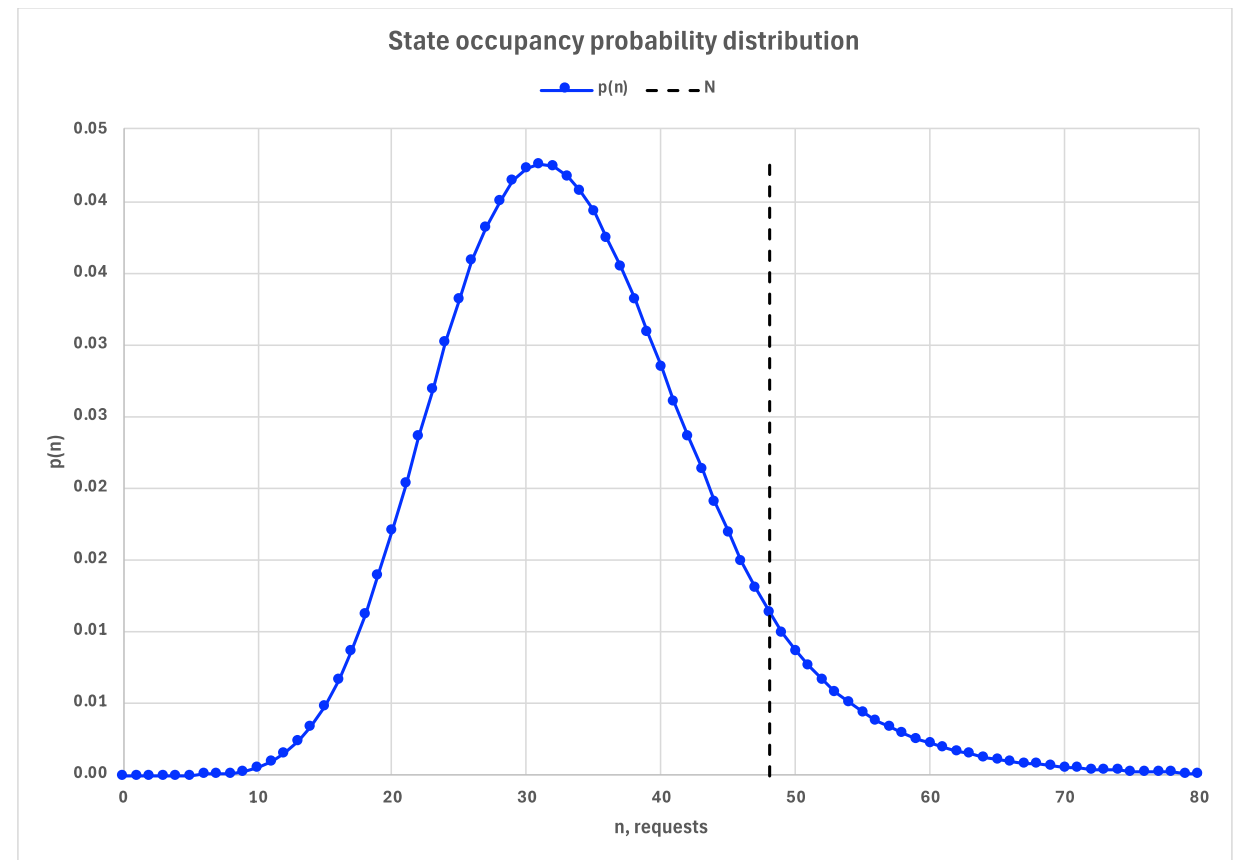
maximum request rate

s.t.

avgWaitTime \leq target

target = 1.0 sec

requestRate = 36.613 req/min



Accelerator data

- accelerator profiles
available, mem size, mem BW, cost
- GPU power
idle and max power, reflection utilization

accelerator-data.json

Model data

- model specs
mem requirements
- model performance on accelerators
token service time parameters
max batch size at tokens

model-data.json

Workload data

- classes of service
request arrivals rate, CoV
model mix
number of tokens (avg, CoV)
ITL & wait SLO constraints

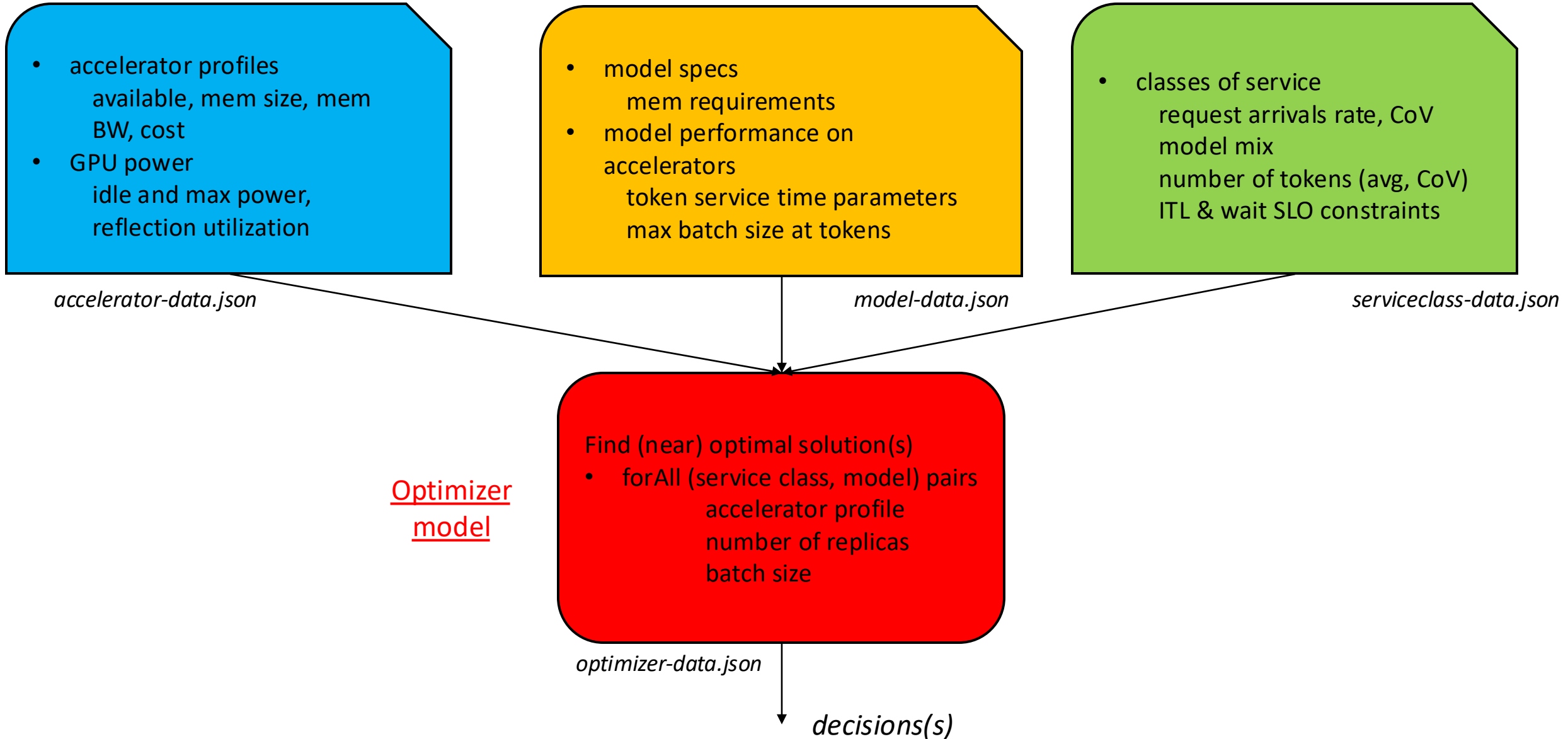
serviceclass-data.json

Optimizer model

- Find (near) optimal solution(s)
- forAll (service class, model) pairs
accelerator profile
number of replicas
batch size

optimizer-data.json

decisions(s)



```

"MI300X": {
  "type": "MI300X",
  "multiplicity": 1,
  "memSize": 192,
  "memBW": 5300,
  "power" : {
    "idle": 220,
    "full": 750,
    "midPower": 650,
    "midUtil": 0.6
  },
  "cost": 65.00
},
"2xA100": {
  "type": "A100",
  "multiplicity": 2,
  "memSize": 160,
  "memBW": 4000,
  "power" : {
    "idle": 300,
    "full": 800,
    "midPower": 640,
    "midUtil": 0.6
  },
  "cost": 80.00
},

```

accelerator-data.json

model-data.json

```

{
  "name": "granite_20b",
  "memSize": 60,
  "perfData": [
    {
      "name": "AIU2",
      "alpha": 297.8,
      "beta": 5,
      "maxBatchSize": 38,
      "atTokens": 512
    },
    {
      "name": "L4",
      "alpha": 198.53,
      "beta": 3.33,
      "maxBatchSize": 7,
      "atTokens": 512
    }
  ],
}

```

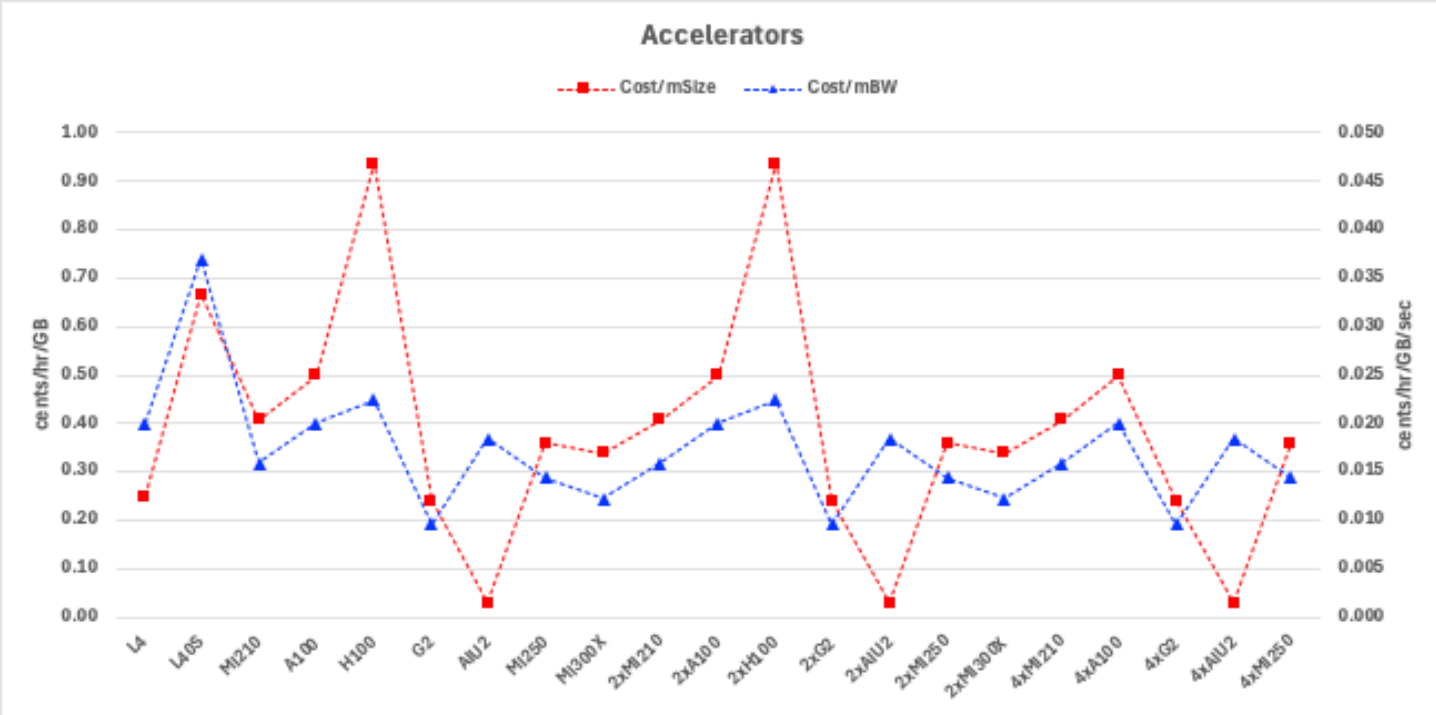
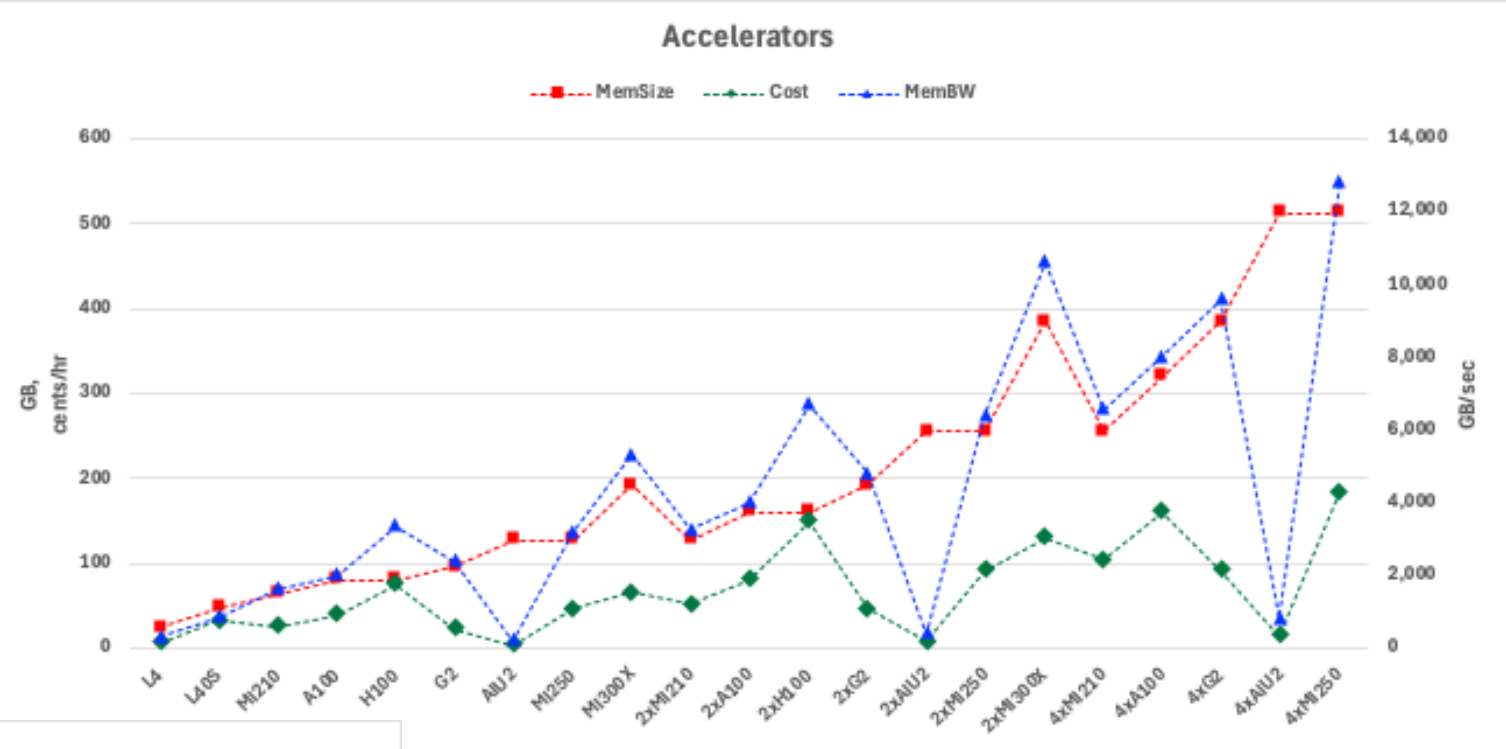
serviceclass-data.json

```

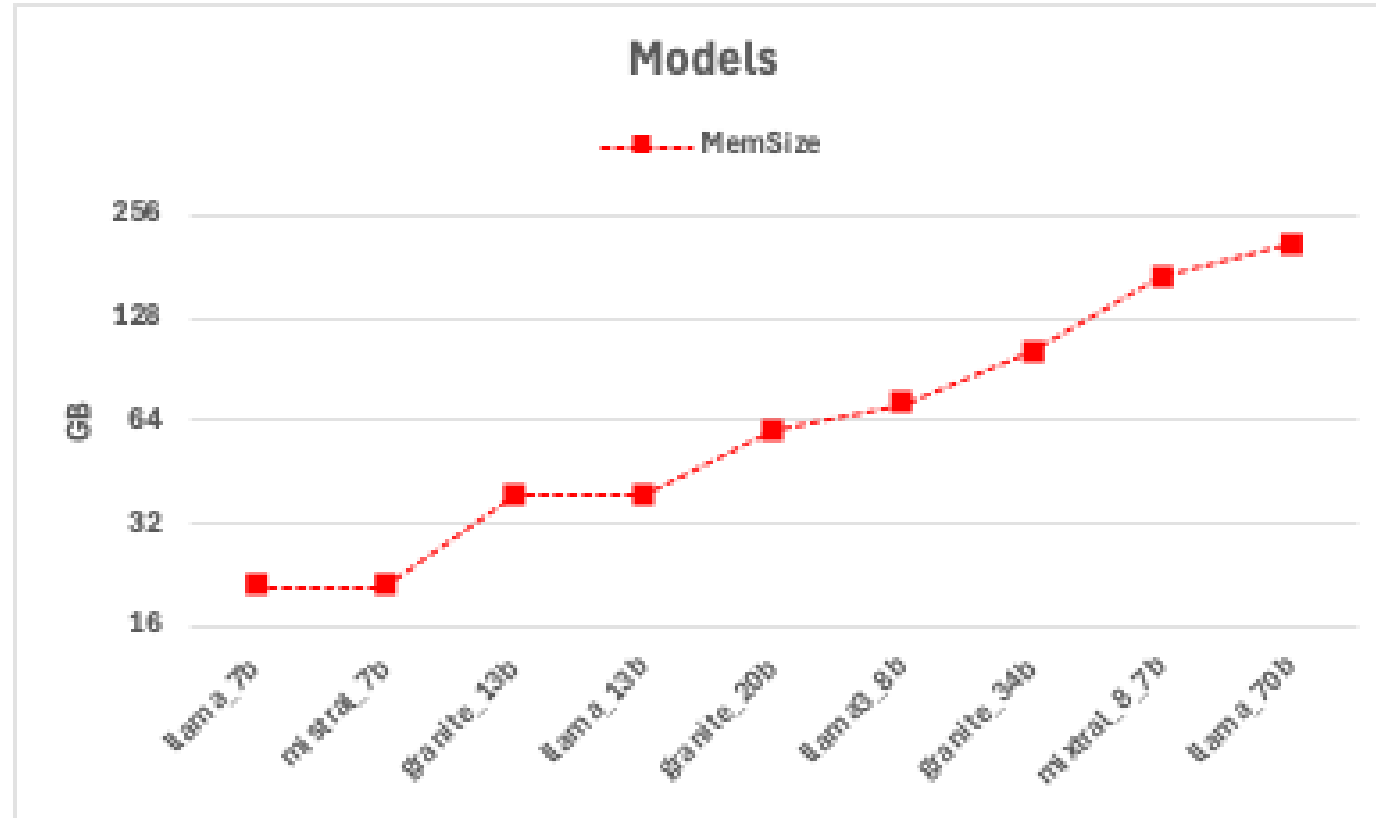
"name": "Premium",
"load": [
  {
    "name": "llama_7b",
    "slo-itl": 40,
    "slo-ttw": 500,
    "arrivalRate": 120,
    "avgLength": 1024,
    "arrivalCOV": 1.5,
    "serviceCOV": 1.5
  },
  {
    "name": "llama_13b",
    "slo-itl": 80,
    "slo-ttw": 500,
    "arrivalRate": 120,
    "avgLength": 1024,
    "arrivalCOV": 1.5,
    "serviceCOV": 1.5
  },
}

```

Accelerators Specs

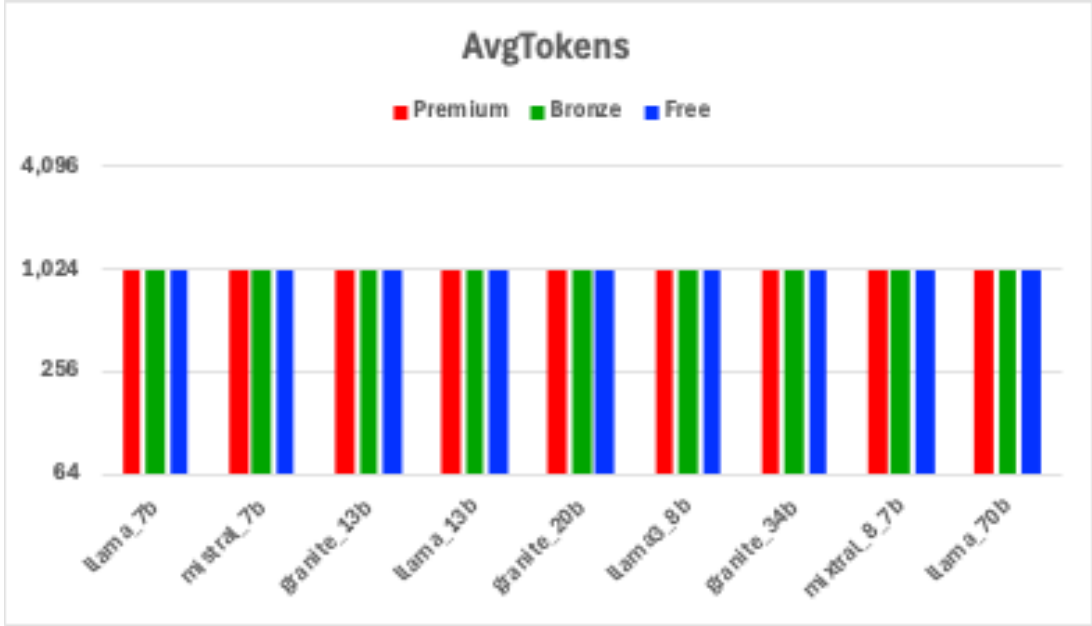
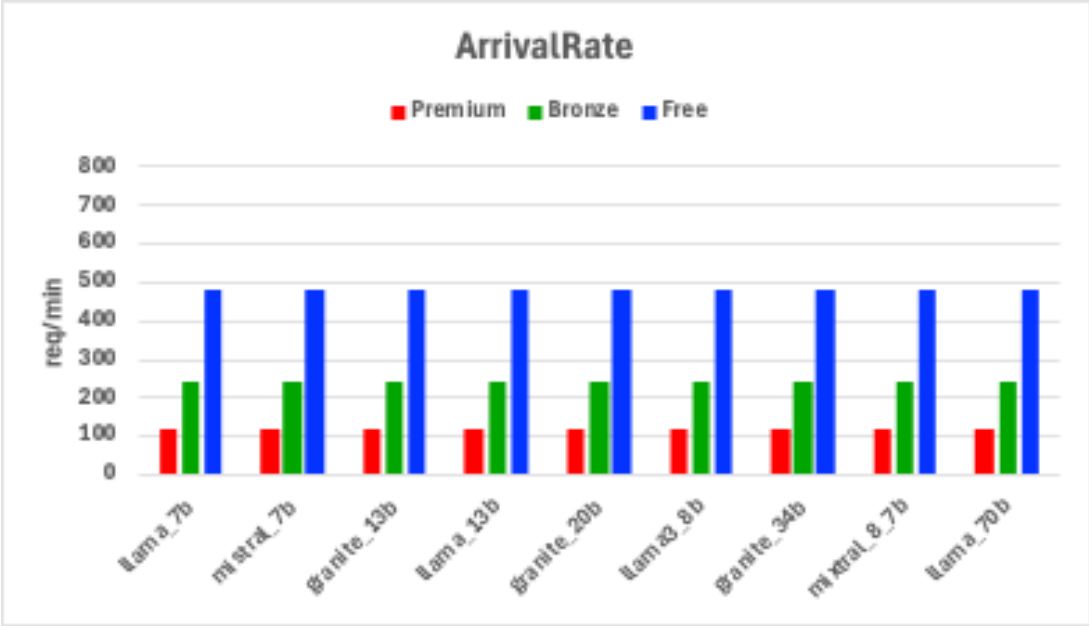


Models Specs

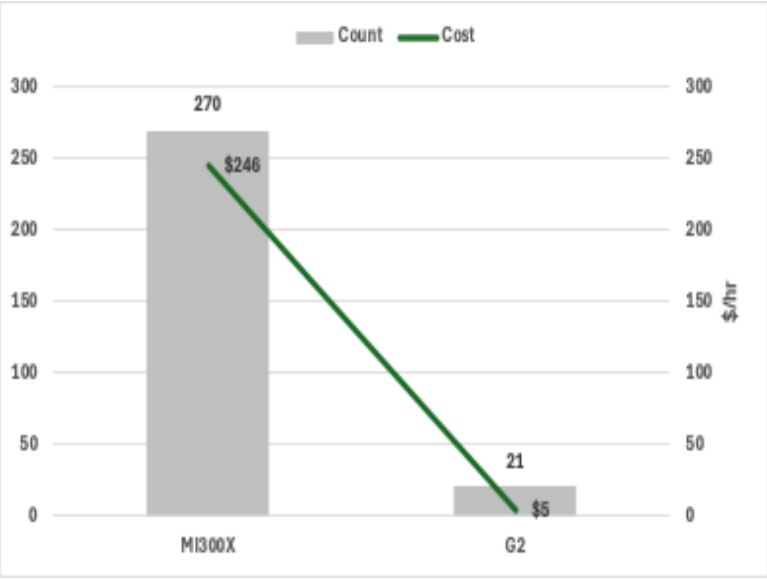


Unlimited accelerators

- capacity planning
- cloud deployment
- separable optimization

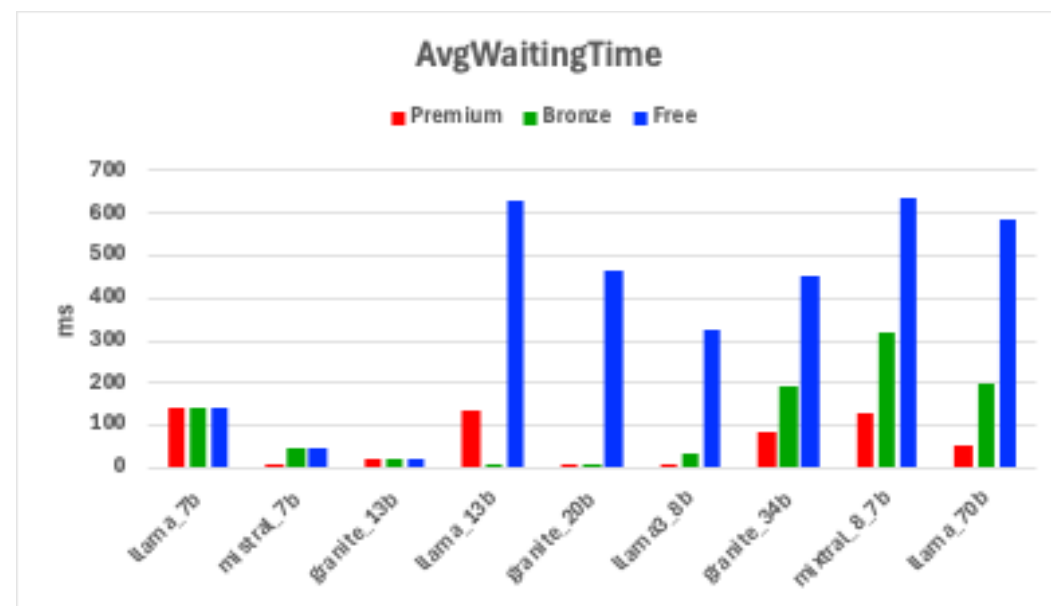
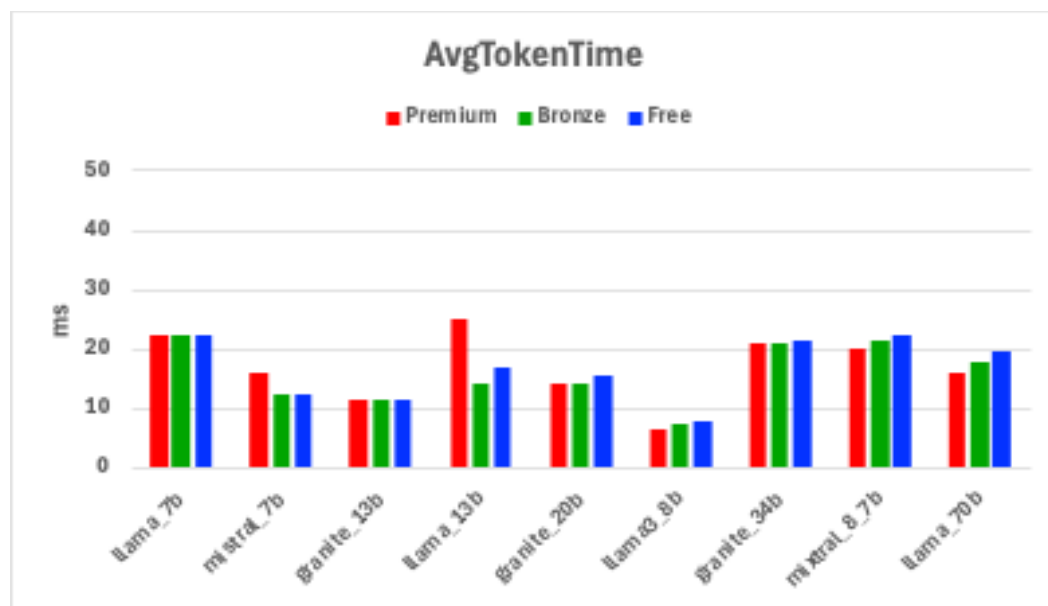
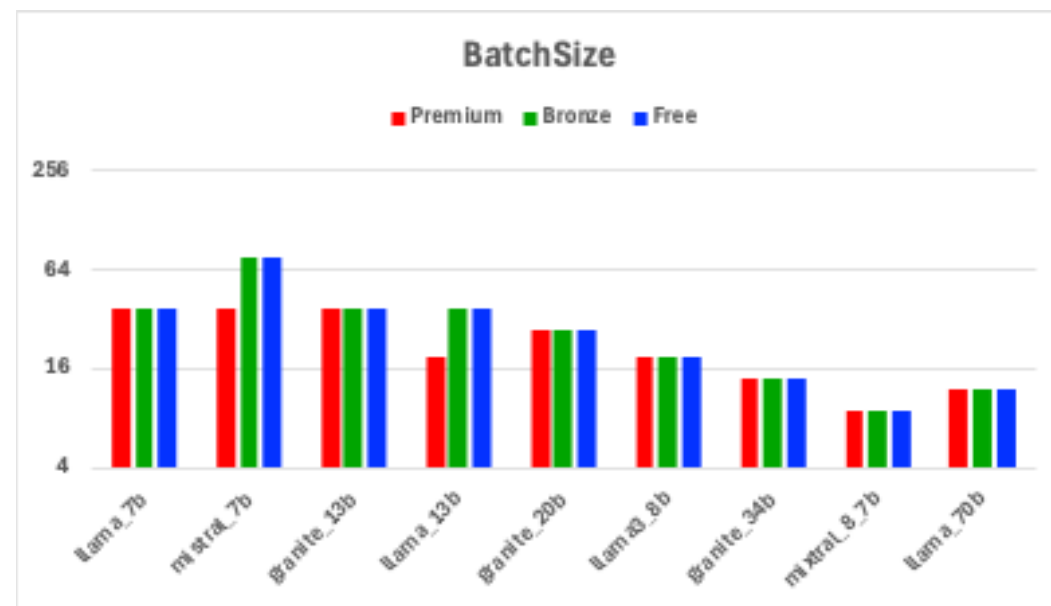
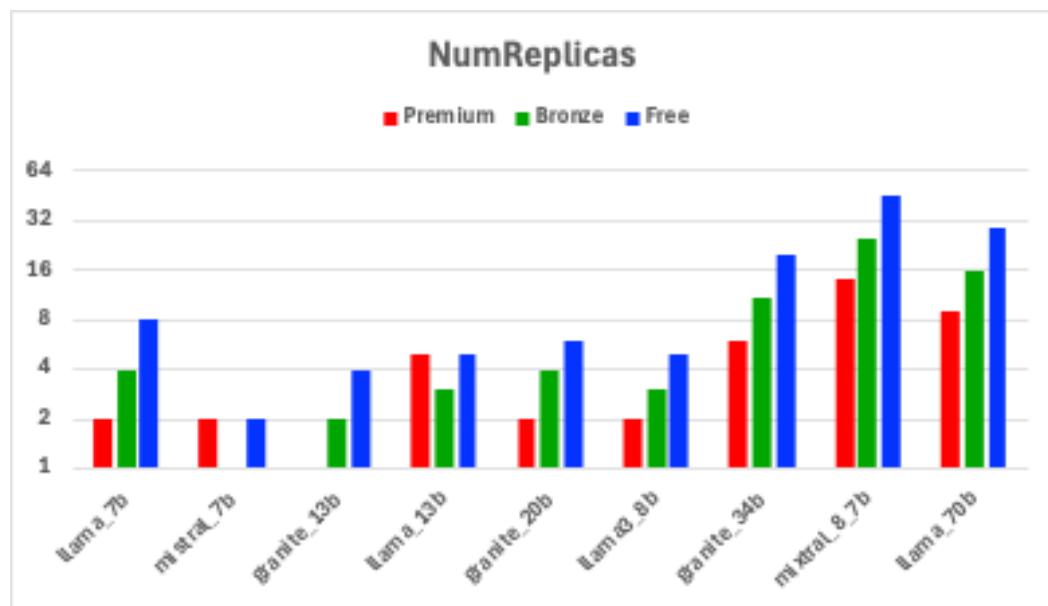


Accelerator			
	Premium	Bronze	Free
llama_7b	G2	G2	G2
mistral_7b	G2	MI300X	MI300X
granite_13b	MI300X	MI300X	MI300X
llama_13b	G2	MI300X	MI300X
granite_20b	MI300X	MI300X	MI300X
llama3_8b	MI300X	MI300X	MI300X
granite_34b	MI300X	MI300X	MI300X
mixtral_8_7b	MI300X	MI300X	MI300X
llama_70b	2xMI300X	2xMI300X	2xMI300X



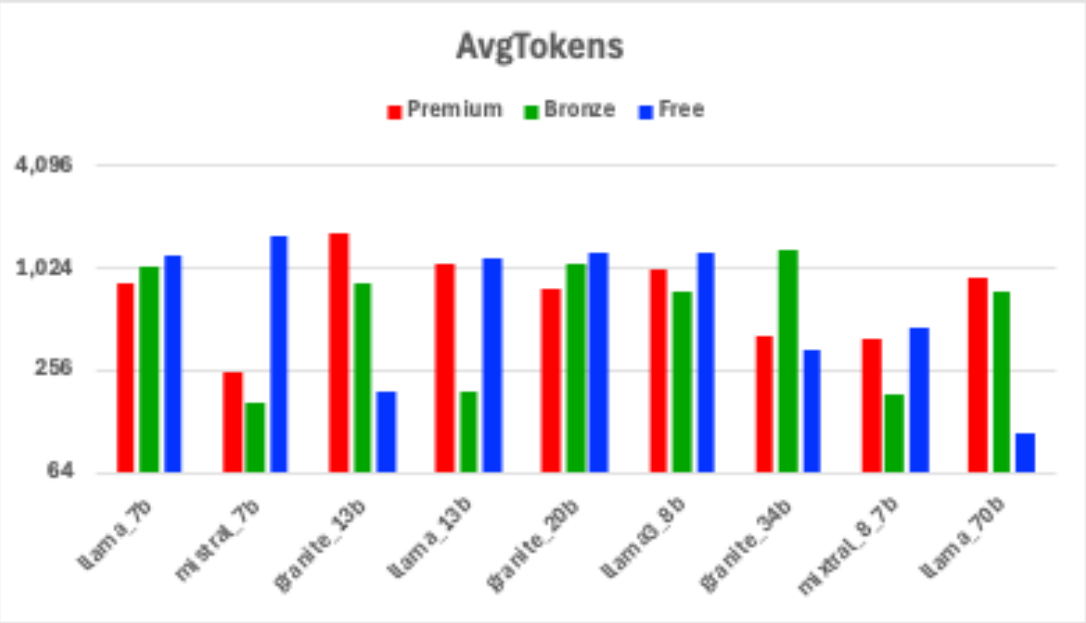
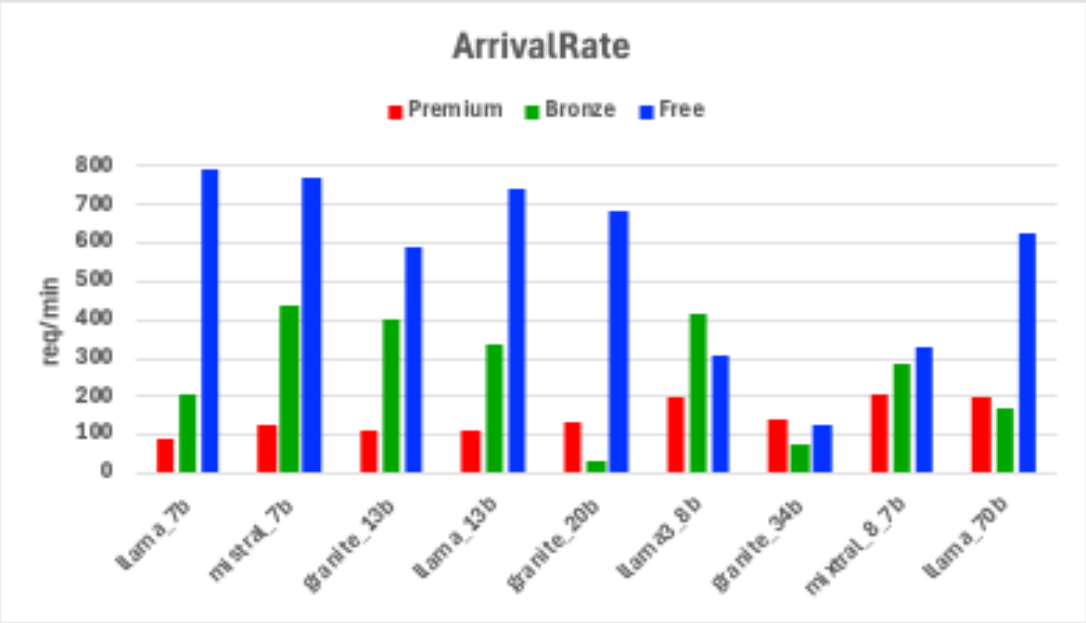
TotalCost
25,053.00



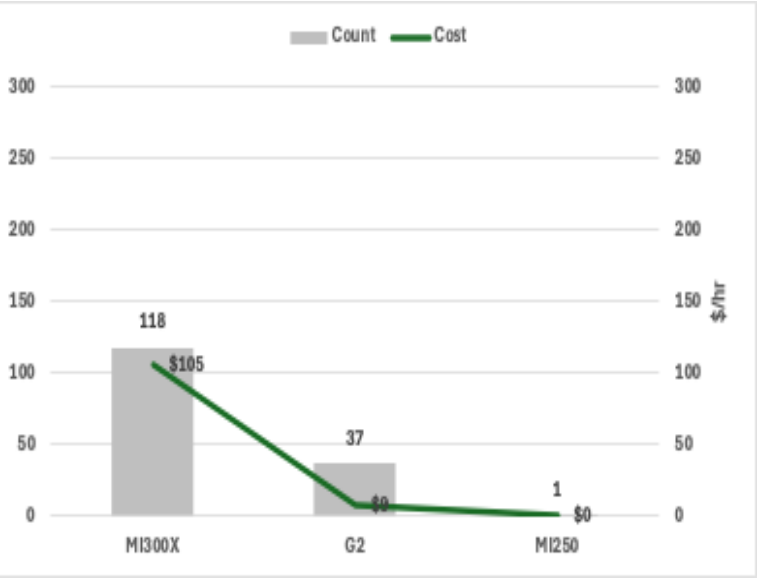


Unlimited accelerators - Dynamic

- change in request rates
- change in request lengths
- change/scale accelerators
- minimize change in accelerators and cost



Accelerator Change			
	Premium	Bronze	Free
llama_7b			
mistral_7b		MI300X->G2	
granite_13b			MI300X->G2
llama_13b		MI300X->MI250	
granite_20b	MI300X->G2		
llama3_8b			
granite_34b			
mixtral_8_7b			
llama_70b			



TotalCost
11,427.00

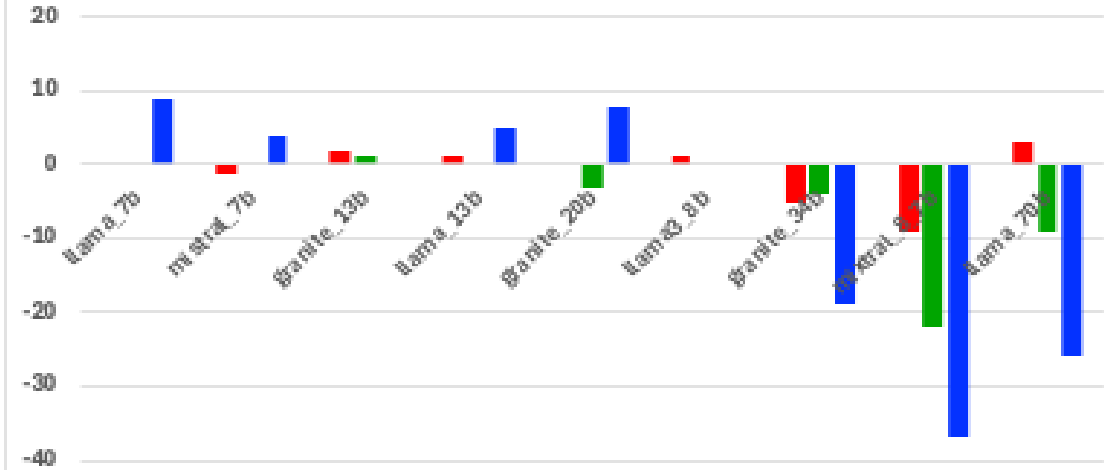


Accelerator Change			
	Premium	Bronze	Free
llama_7b			
mistral_7b		MI300X->G2	
granite_13b			MI300X->G2
llama_13b		MI300X->MI250	
granite_20b	MI300X->G2		
llama3_8b			
granite_34b			
mixtral_8_7b			
llama_70b			

Scale			
	Premium	Bronze	Free
llama_7b	0	0	9
mistral_7b	-1		4
granite_13b	2	1	
llama_13b	1		5
granite_20b		-3	8
llama3_8b	1	0	0
granite_34b	-5	-4	-19
mixtral_8_7b	-9	-22	-37
llama_70b	3	-9	-26

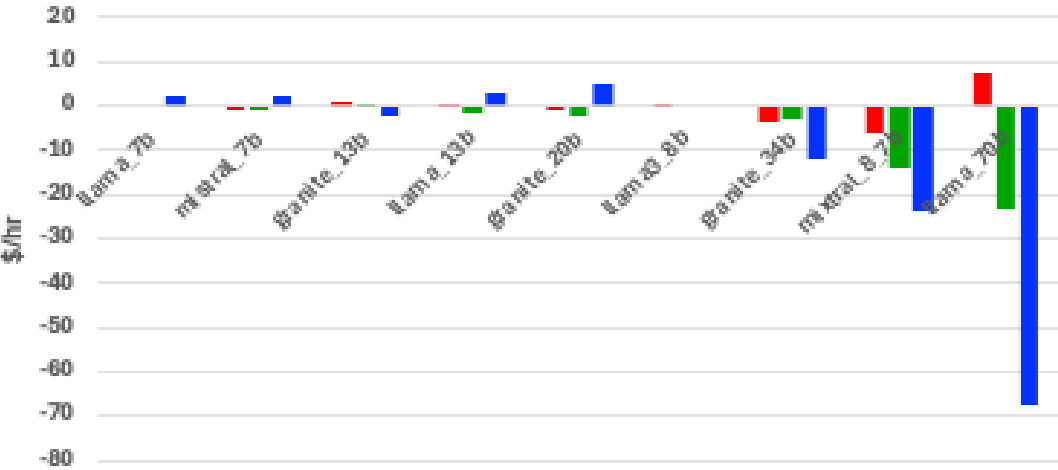
Scaling NumReplicas

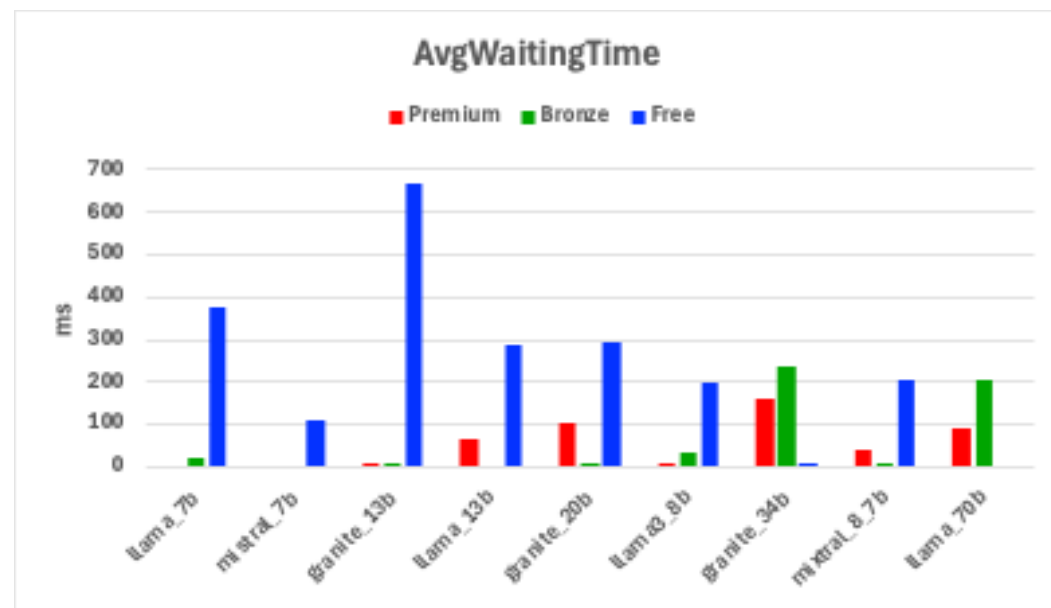
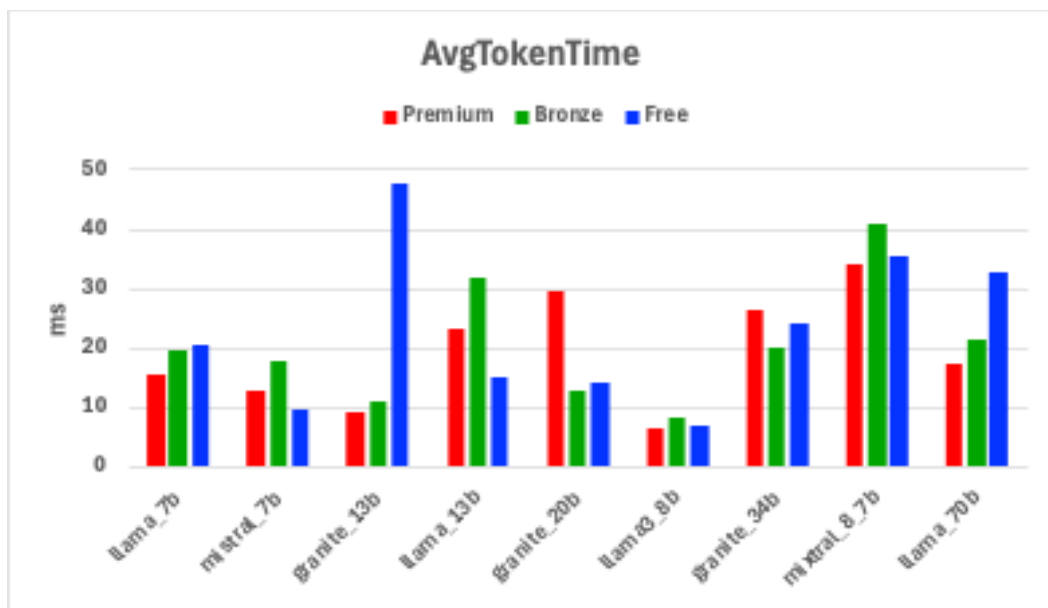
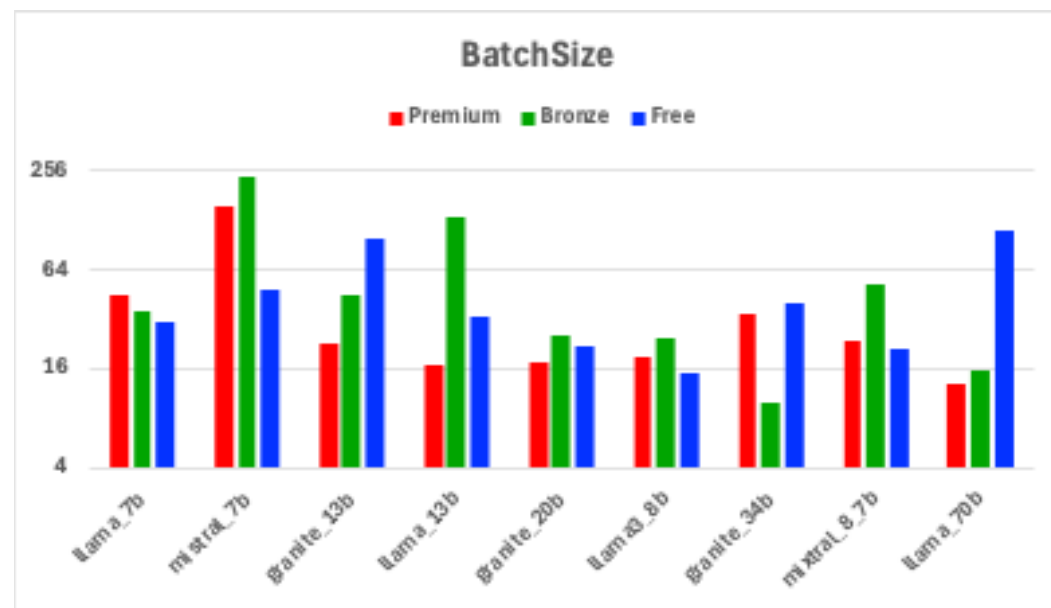
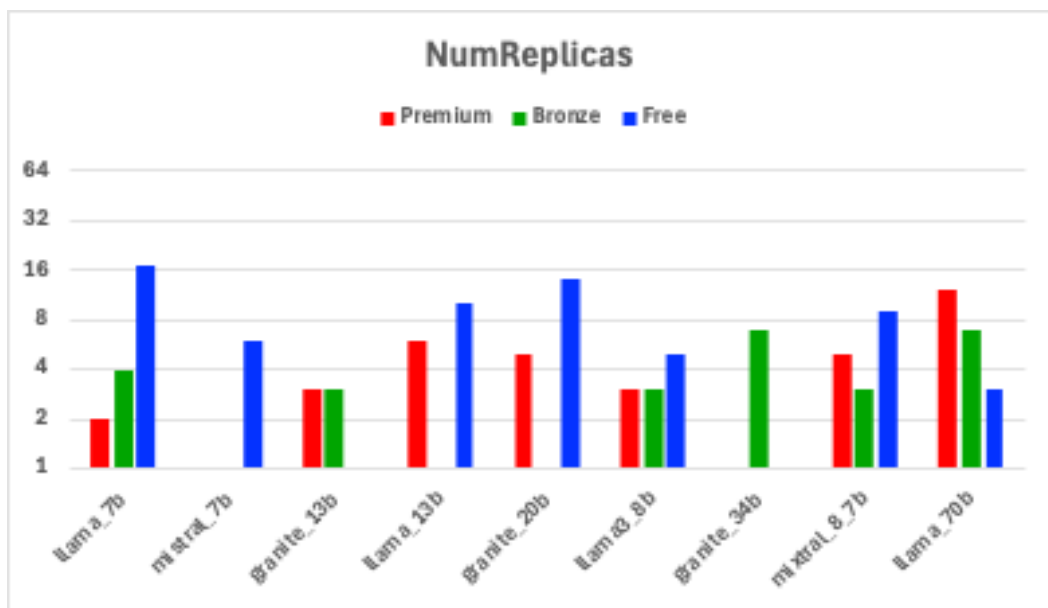
Premium Bronze Free

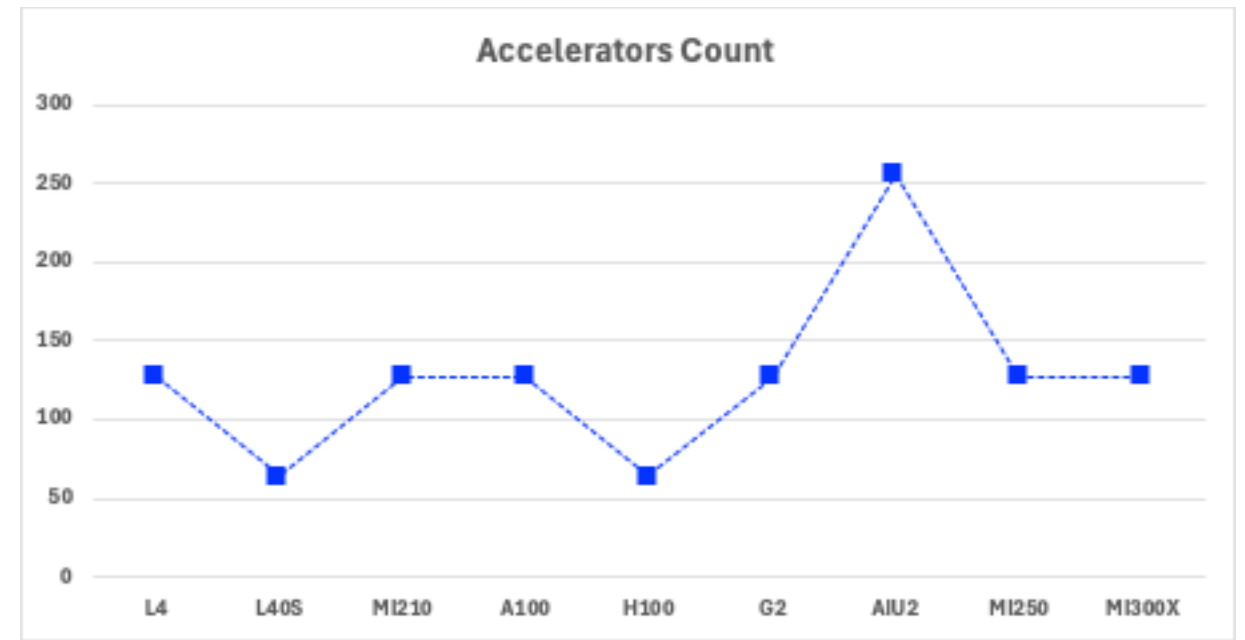


Cost differential

Premium Bronze Free

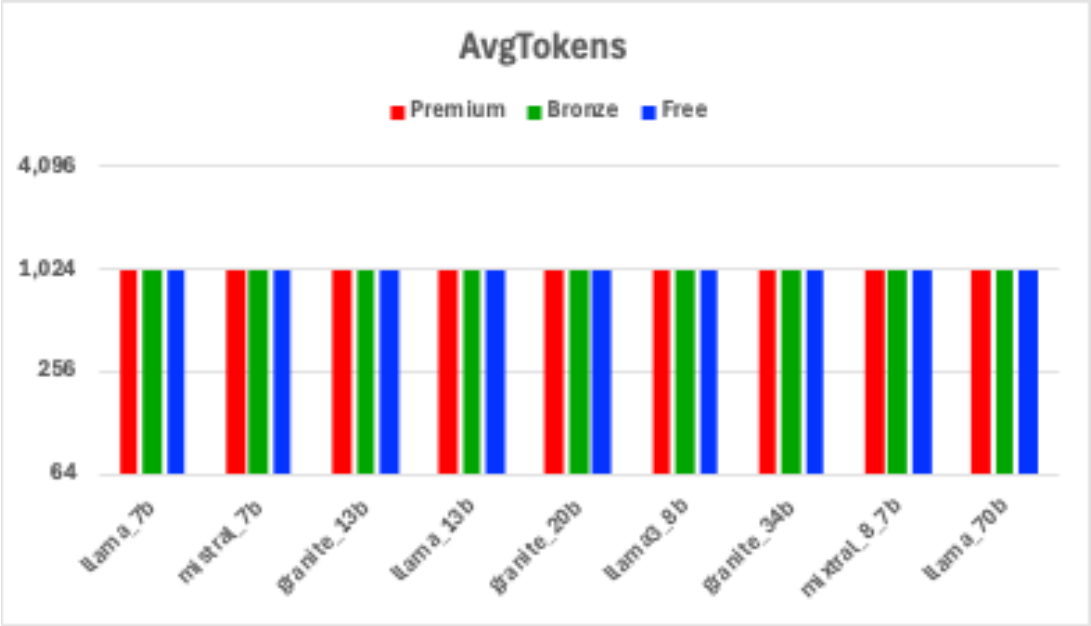
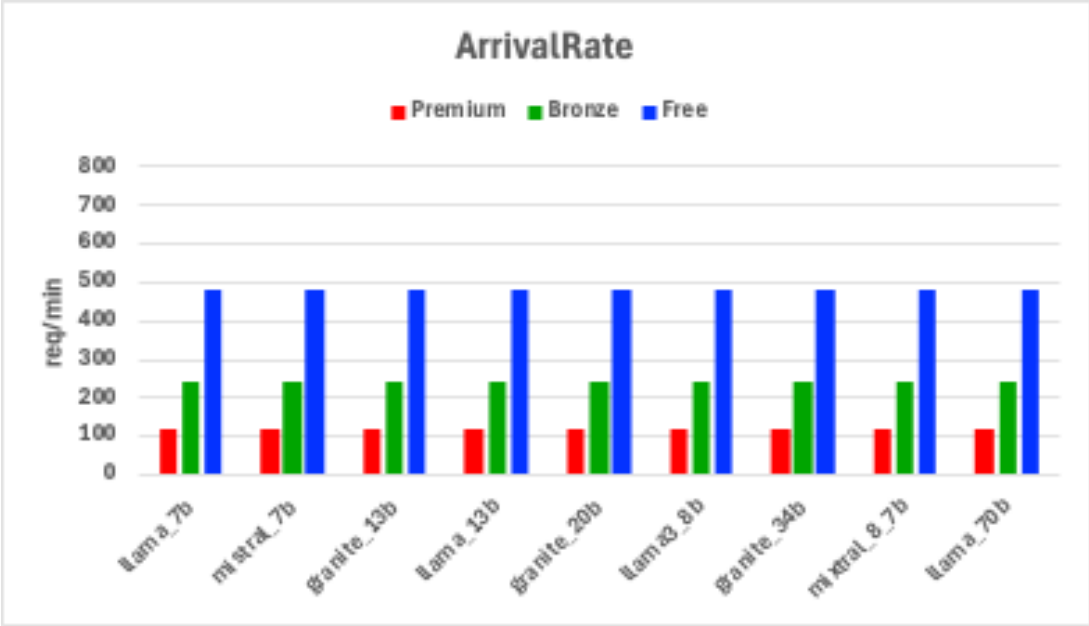




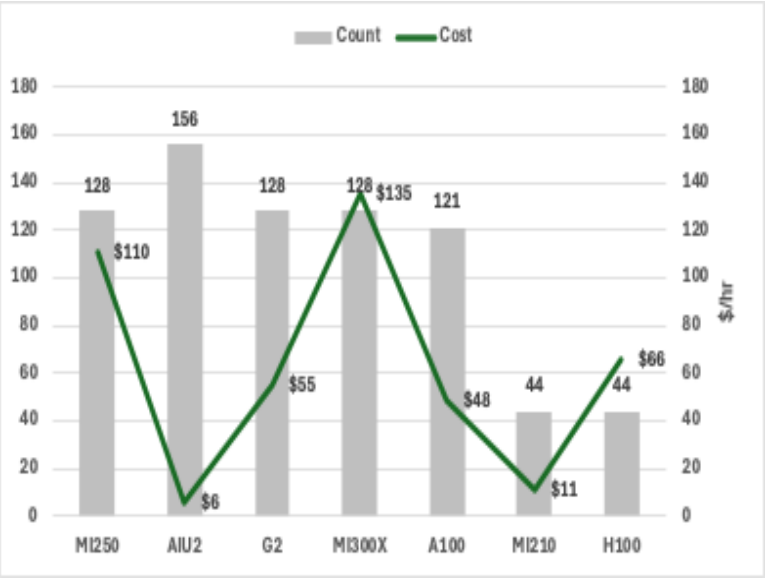


Limited accelerators

- cluster deployment
- greedy optimization

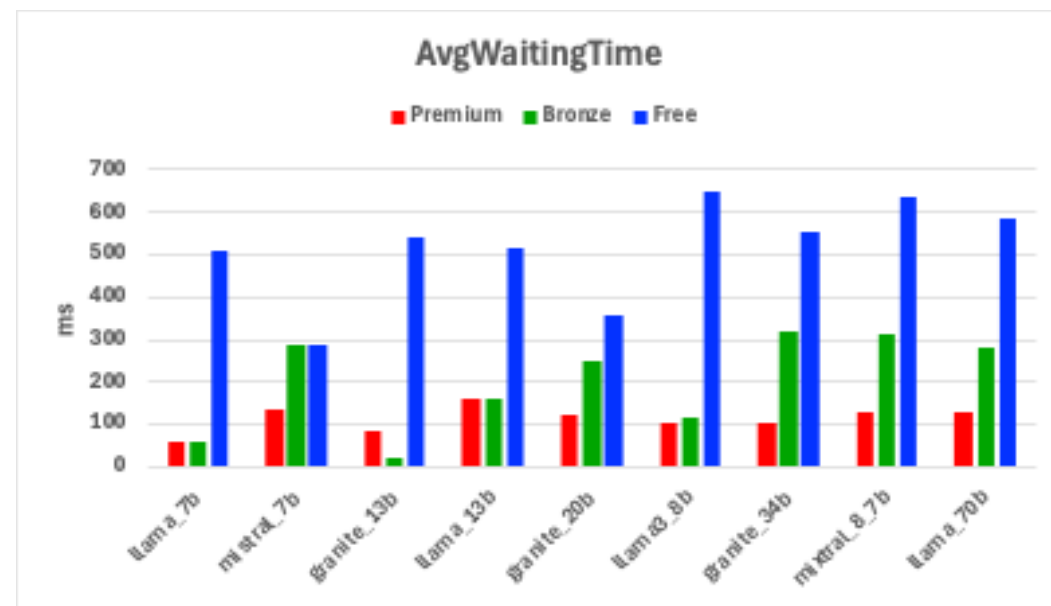
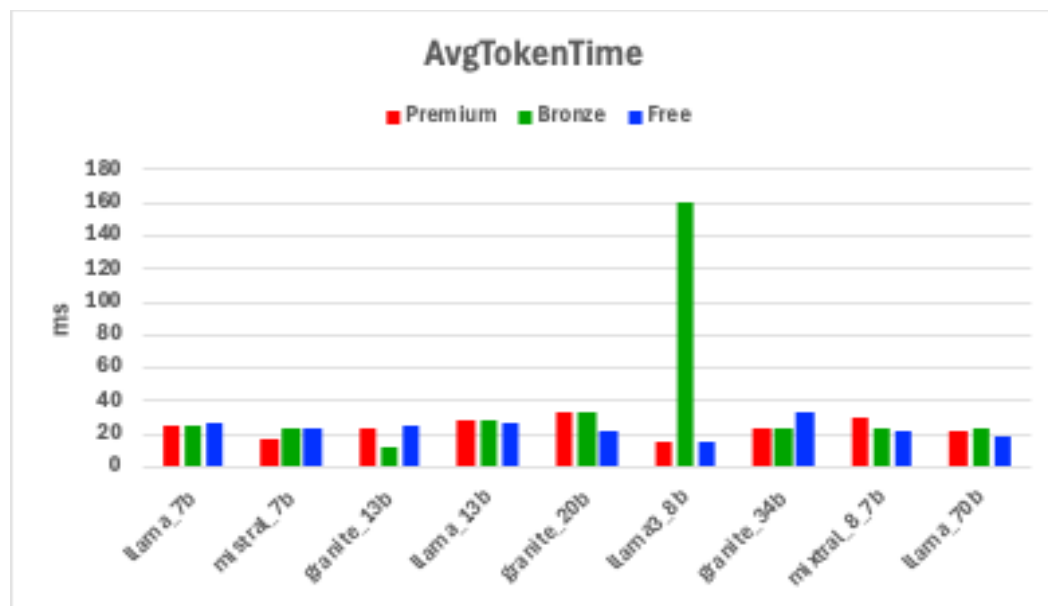
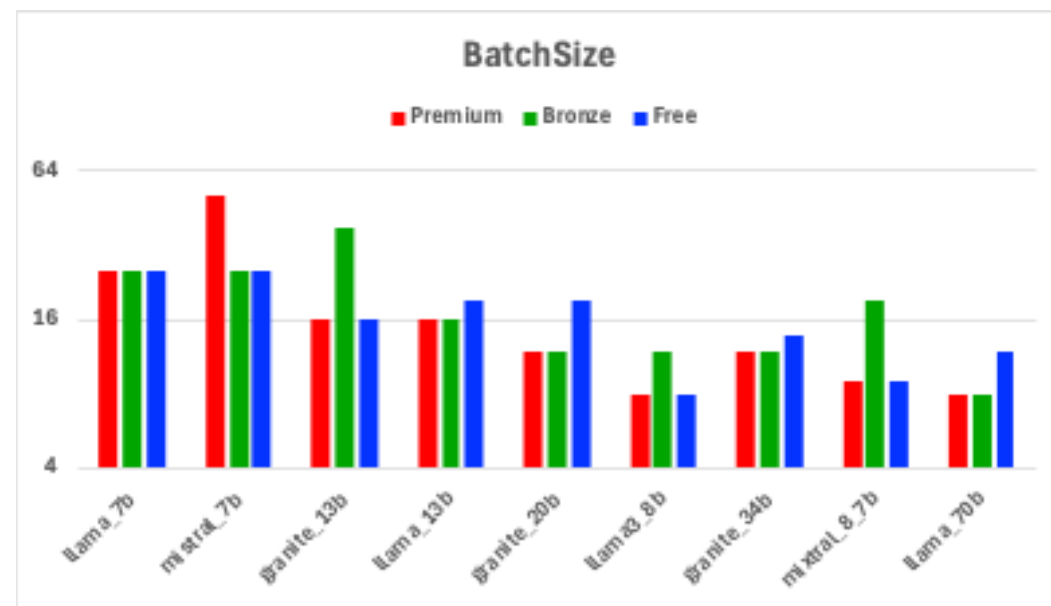
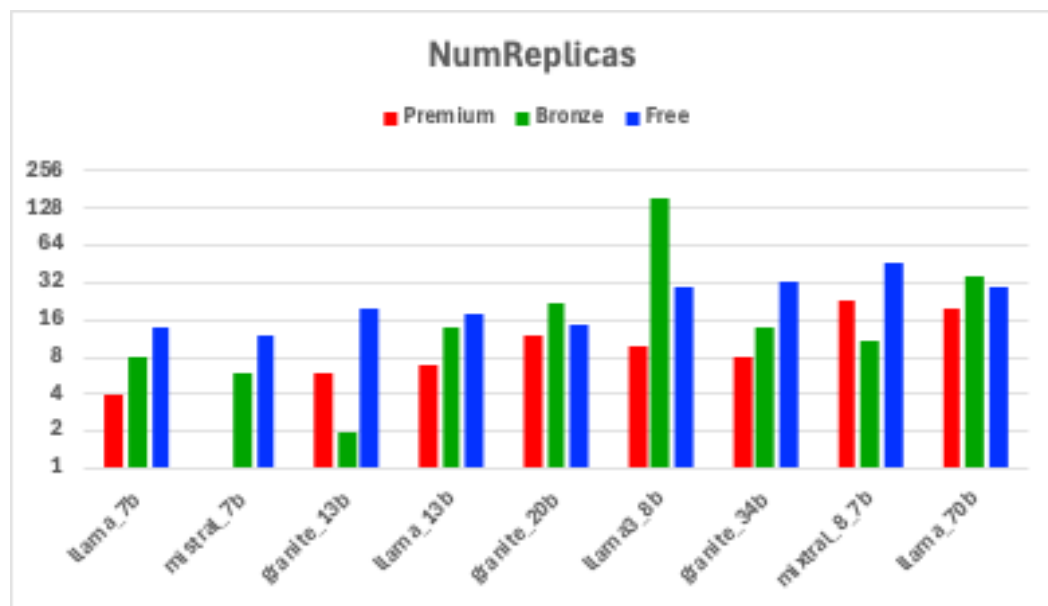


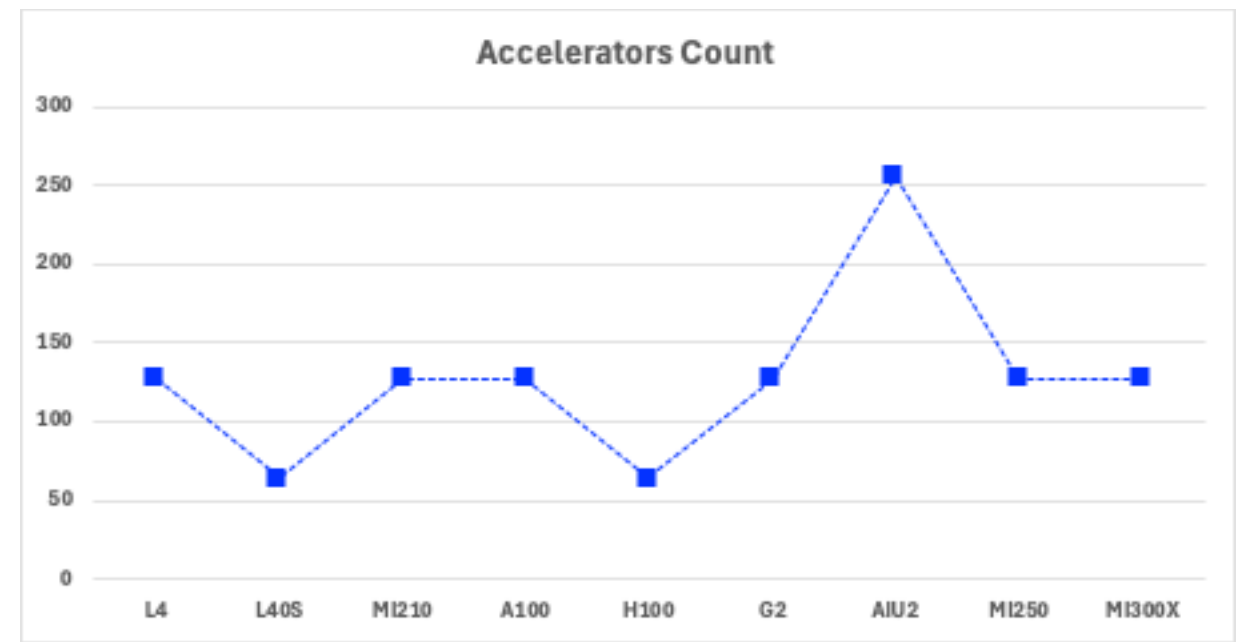
Accelerator			
	Premium	Bronze	Free
llama_7b	MI210	MI210	MI210
mistral_7b	MI250	MI210	MI210
granite_13b	A100	MI300X	A100
llama_13b	A100	A100	G2
granite_20b	A100	A100	MI250
llama3_8b	A100	AIU2	A100
granite_34b	2xH100	2xH100	2xG2
mixtral_8_7b	2xG2	2xMI300X	MI300X
llama_70b	2xMI250	2xMI250	2xMI300X



TotalCost
43,203.00

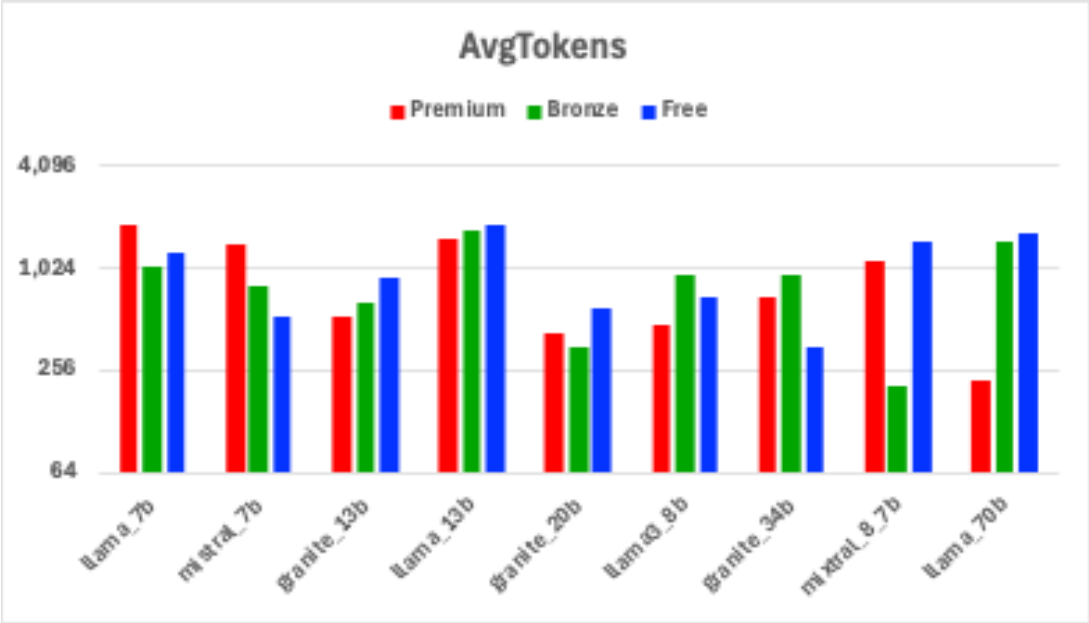




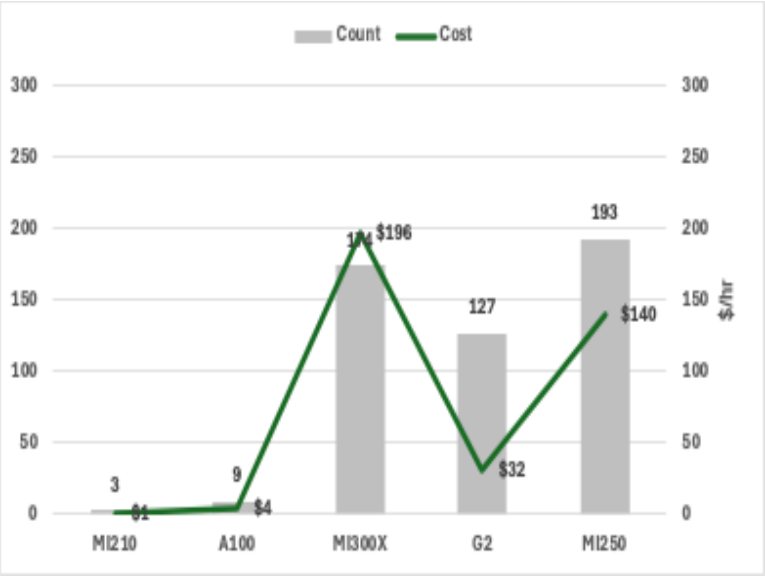


Limited accelerators - Dynamic

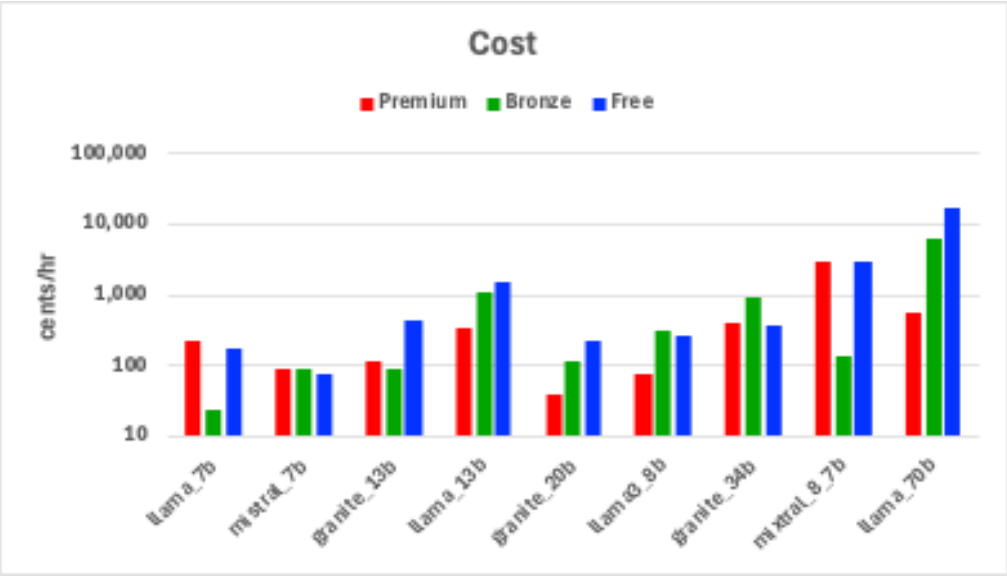
- change in request rates
- change in request lengths
- change/scale accelerators
- minimize change in accelerators and cost



Accelerator Change			
	Premium	Bronze	Free
llama_7b	MI210->MI250	MI210->G2	MI210->MI250
mistral_7b		MI210->MI250	
granite_13b		MI300X->MI250	A100->G2
llama_13b	A100->G2	A100->MI250	
granite_20b			
llama3_8b		A100->MI250	A100->G2
granite_34b	2xH100->MI250	2xH100->MI250	
mixtral_8_7b	2xG2->2xMI250	2xMI300X->G2	
llama_70b			



TotalCost
37,272.00

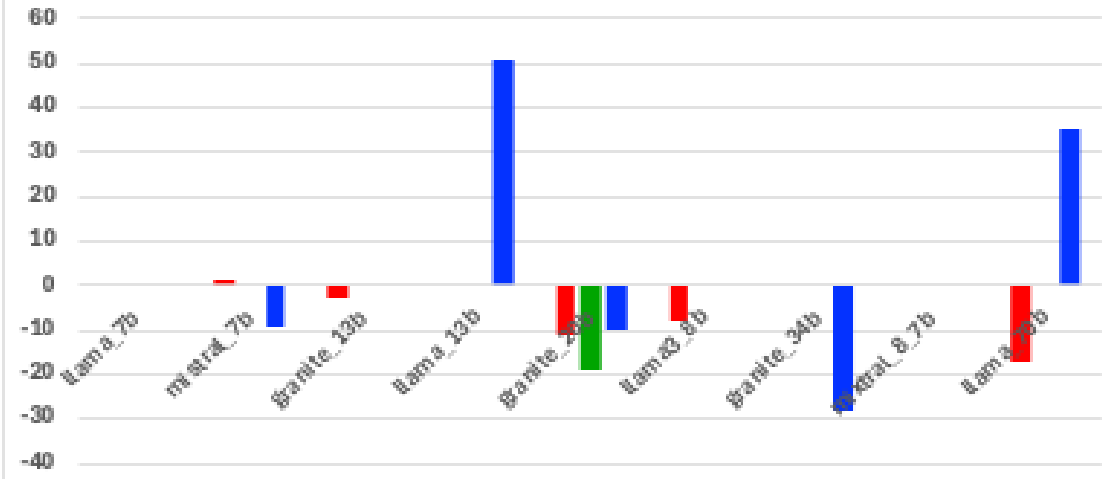


Accelerator Change			
	Premium	Bronze	Free
llama_7b	MI210->MI250	MI210->G2	MI210->MI250
mistral_7b		MI210->MI250	
granite_13b		MI300X->MI250	A100->G2
llama_13b	A100->G2	A100->MI250	
granite_20b			
llama3_8b		A1U2->MI250	A100->G2
granite_34b	2xH100->MI250	2xH100->MI250	
mixtral_8_7b	2xG2->2xMI250	2xMI300X->G2	
llama_70b			

Scale			
	Premium	Bronze	Free
llama_7b			
mistral_7b	1		-9
granite_13b	-3		
llama_13b			51
granite_20b	-11	-19	-10
llama3_8b	-8		
granite_34b			-28
mixtral_8_7b			0
llama_70b	-17	0	35

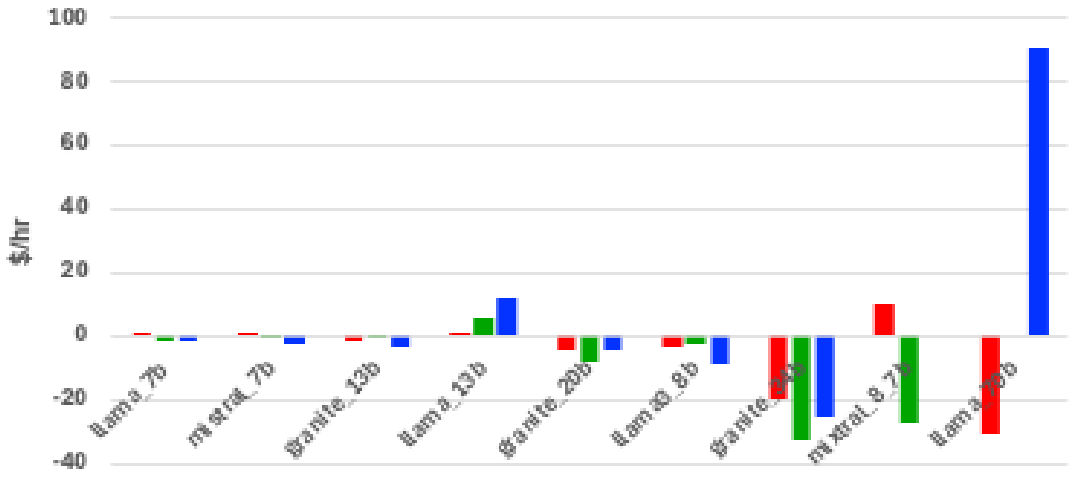
Scaling NumReplicas

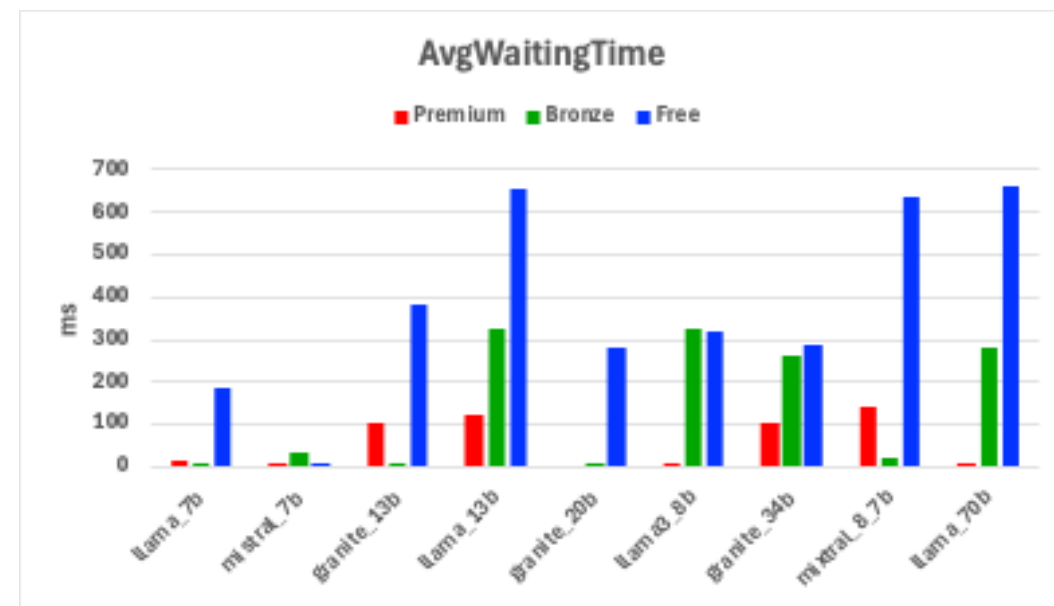
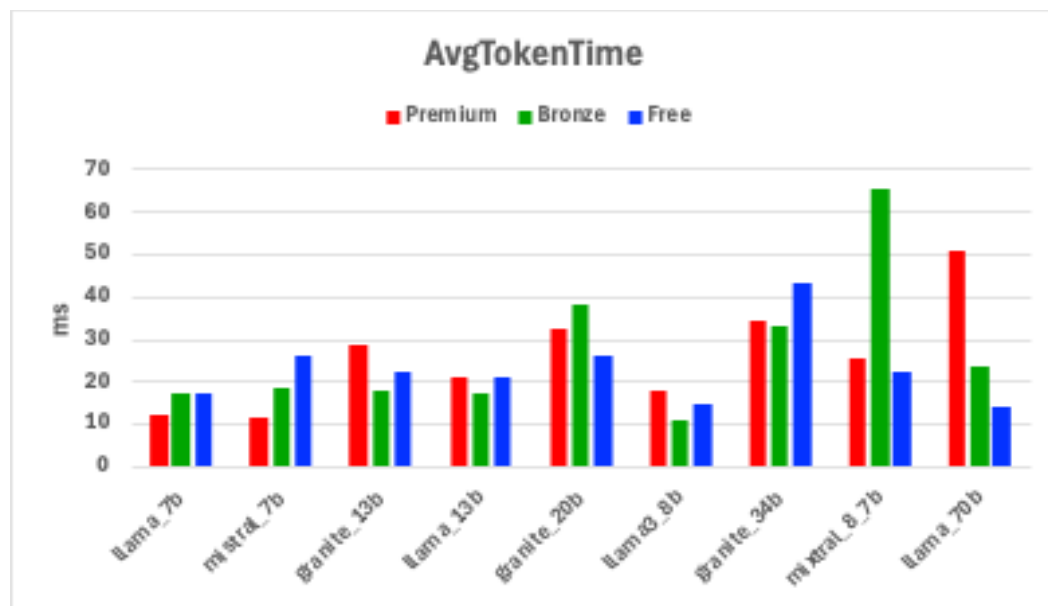
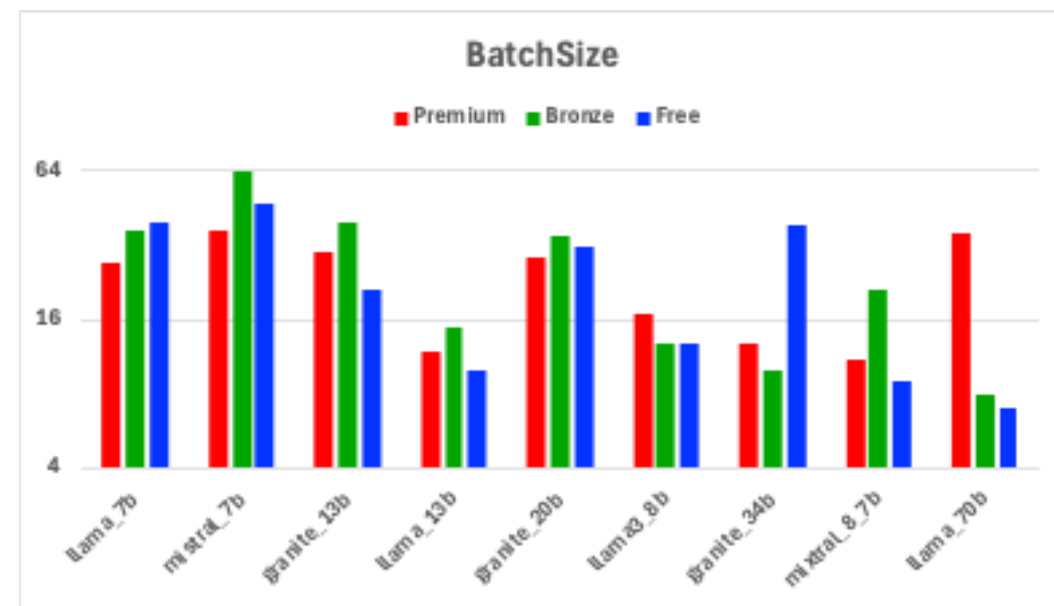
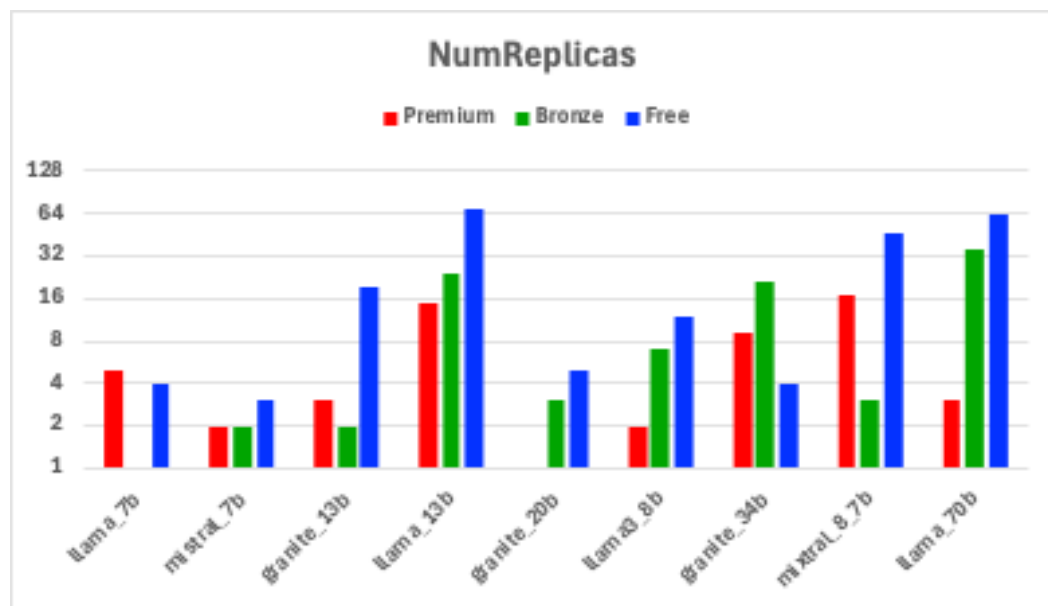
Premium Bronze Free



Cost differential

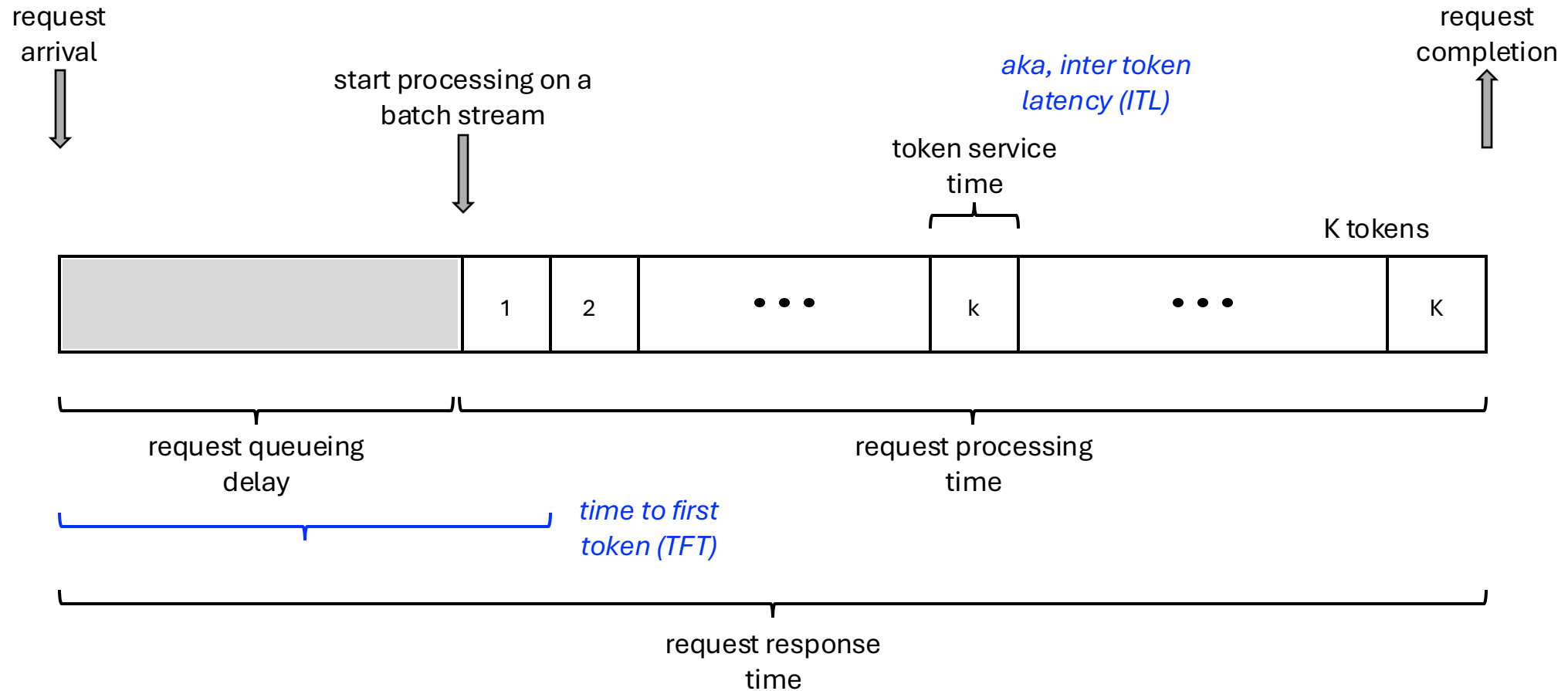
Premium Bronze Free





Backup

Request timing & definitions



amortize delay
over all tokens

$$\text{token delay} = \frac{\text{request response time}}{K}$$

$$\text{token delay} \cong \text{ITL} + \frac{\text{TFT}}{K}$$

Objectives



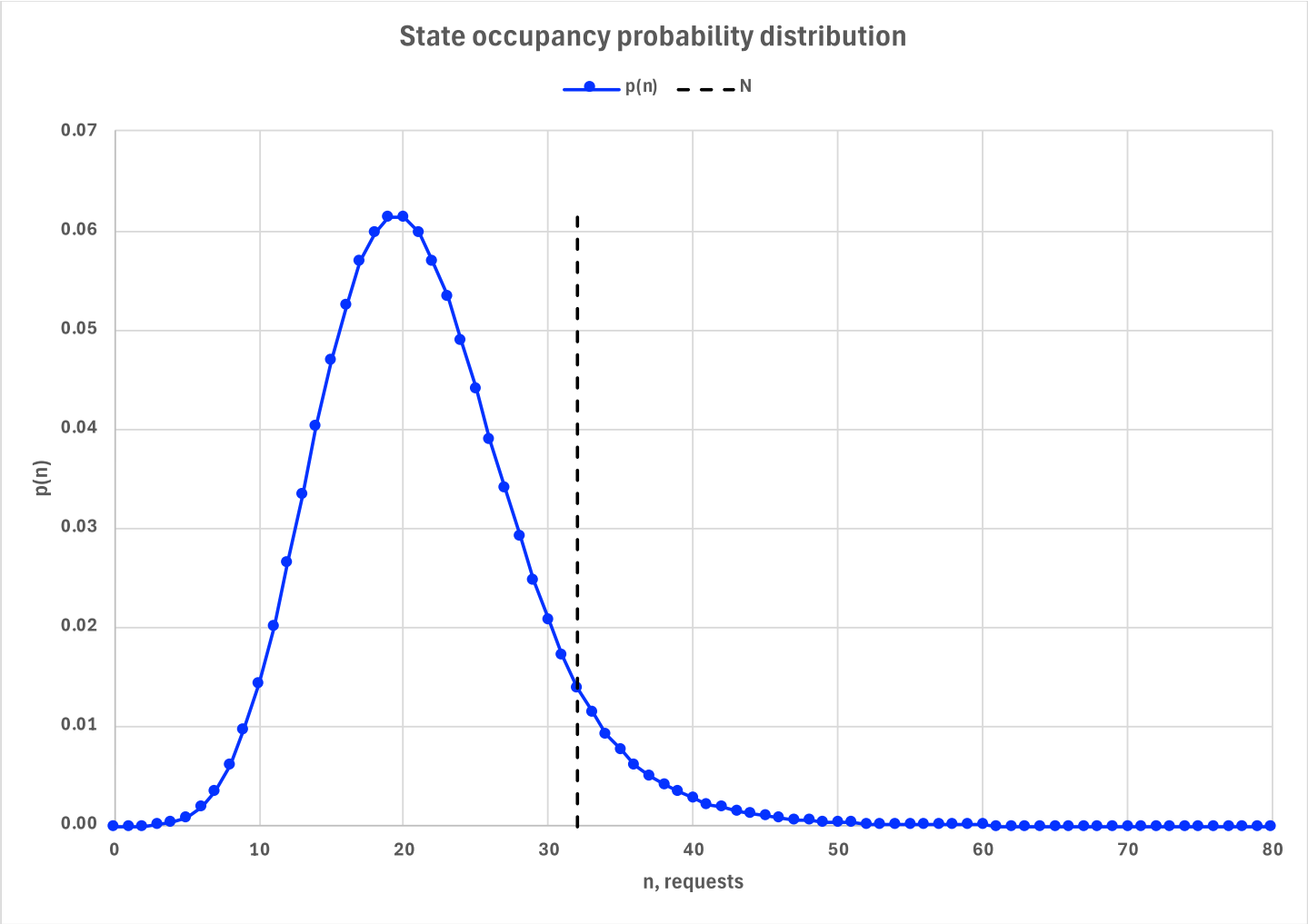
- Optimize operational cost/performance/power
 - assignment of accelerator types
 - model replicas
 - model batch decision
 - meet SLOs
 - min cost and power
- Capacity planning and/or model placement & accelerator allocation



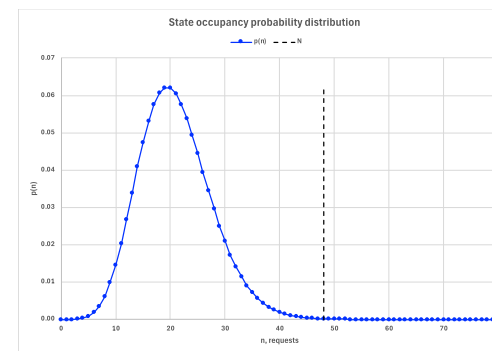
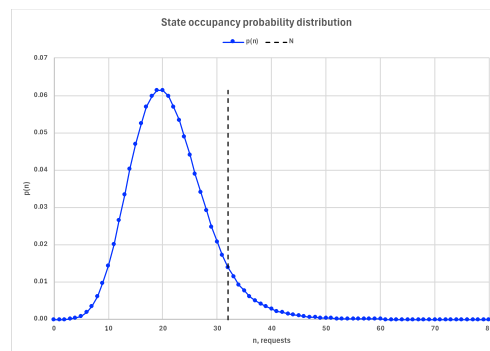
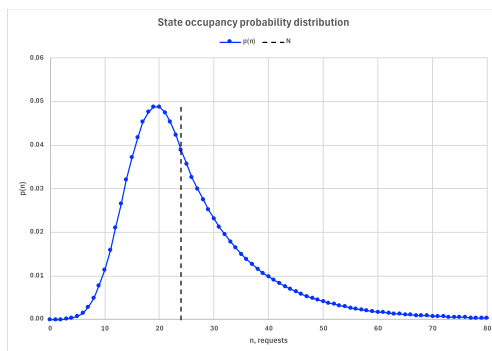
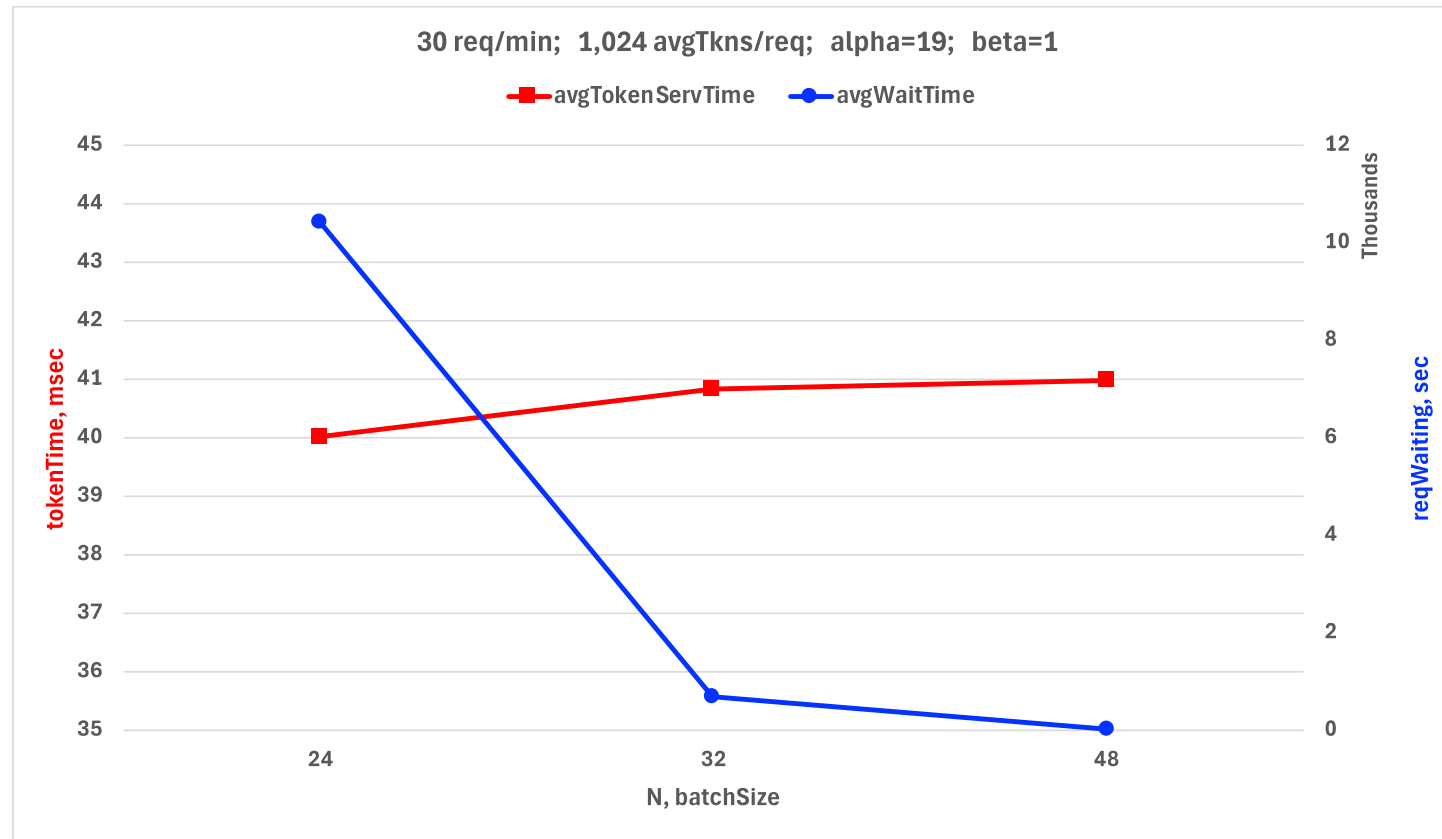
- Inference model optimization
- Improve model accuracy, efficiency, compute/memory requirements, ...
- Explore mechanisms for request queueing, balancing, ...
- Model sharing, adapters, multiplexing, ...
- GPU slicing mechanisms, DRA, migration, ...

sample case

requestRate	30 /min	p(0)	5.7982E-07	avgNumSystem	21.24
lambda	0.0005 /msec	maxN	200	avgNumServer	20.90
K	1024	pFull	2.0382E-17	avgNumQueue	0.34
N	32	effecTput	30.00000 req/m	utilization	0.653
alpha	19		0.00050 /msec	avgRespTime	42,481
beta	1	P(N)	0.93803	avgSrvTime	41,808
lambda*K	0.512			avgWaitTime	674
				avgTokenServTime	40.8



impact of batch size



ITL limiting

requestRate	35.200 /min	p(0)	1.0542E-08	avgNumSystem	30.23
lambda	0.00059 /msec	maxN	200	avgNumServer	30.04
K	1024	pFull	1.4442E-14	avgNumQueue	0.20
N	48	effecTput	35.20031 req/m	utilization	0.626
alpha	19		0.00059 /msec	avgRespTime	51,533
beta	1	P(N)	0.96847	avgSrvTime	51,200
lambda*K	0.600752			avgWaitTime	333
				avgTokenServTime	50.0

Find

maximum request rate

s.t.

avgTokenServTime \leq target

target = 50 msec

requestRate = 35.2 req/min

