Pauses/throughput/latency/aggressiveness

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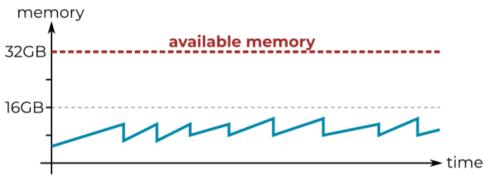
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- and that's what this lesson (an the whole course) is about 😁

An obvious metric - how much memory is being consumed when running **the same** program:

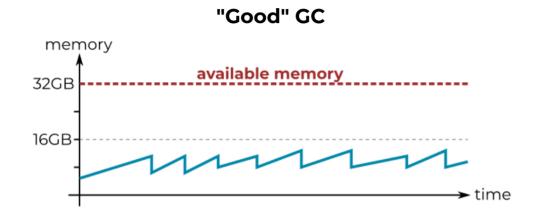
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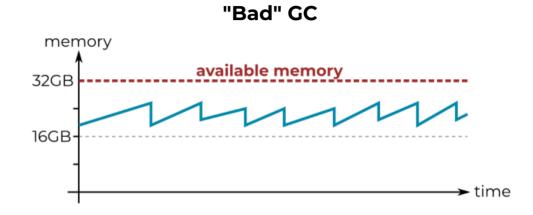


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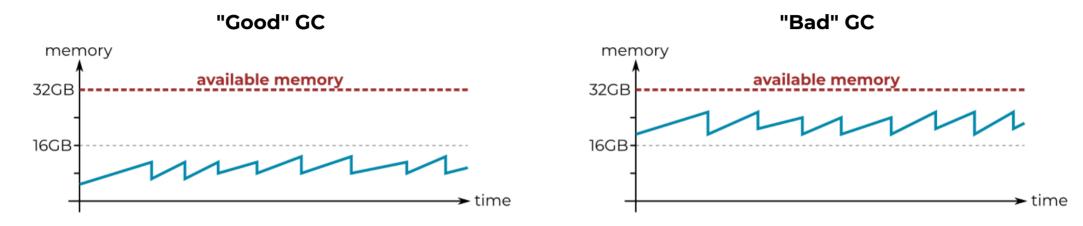


It keeps memory usage constantly small



It keeps memory usage significantly bigger all the time

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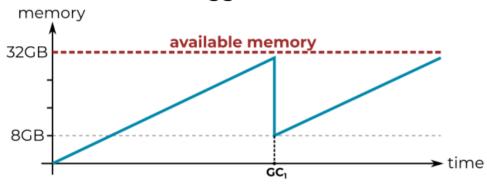
It keeps memory usage significantly bigger all the time

- but... it is not so easy to say "good" or "bad" what's **the cost** of smaller memory usage? CPU? application slowdown?
- if and only if all is the same except memory usage we have better or worse GC

More often, we talk about GC's **aggressiveness** - how often and how much memory it is reclaiming:

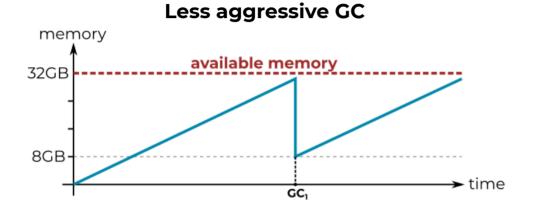
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Less aggressive GC

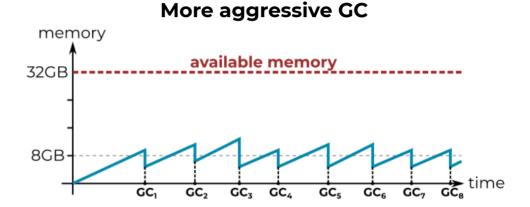


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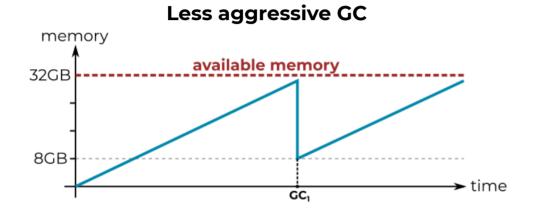


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- memory footprint is significantly smaller

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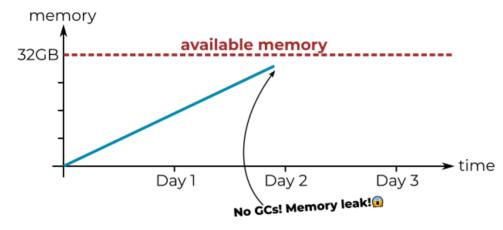
- does GC very often reclaiming memory in small batches (it is rather fast)
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Aggressiveness influences other metrics - the less frequent GCs, the bigger overhead they introduce (longer pauses, bigger CPU spikes).

Aggressiveness is a pretty frequent source of confusion:

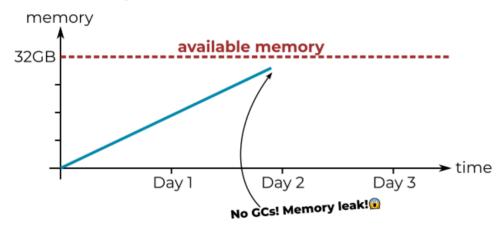
Aggressiveness is a pretty frequent source of confusion:

1. We see a lot of memory growing up to GBs (and sometimes even no GCs at all) and treat it as a memory leak!

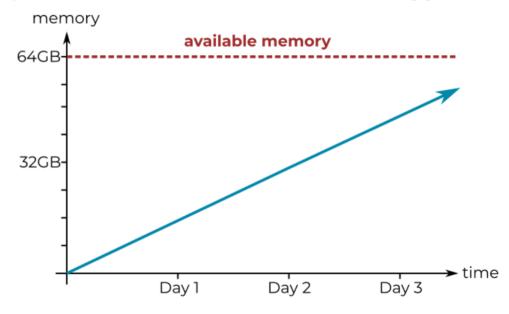


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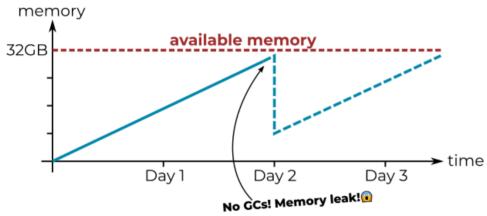
2. In fear, we add more RAM to avoid killing the process and "the leak" becomes even bigger!



Aggressiveness is a pretty frequent source of confusion (cont.):

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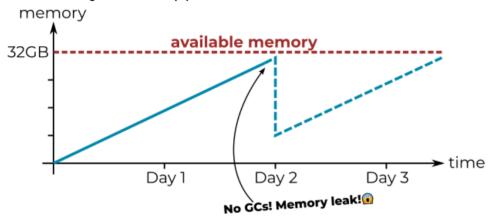
3. If we waited a little longer, the GC would be eventually called (*):



^{*} unless we really have a memory leak, not aggressiveness misunderstanding

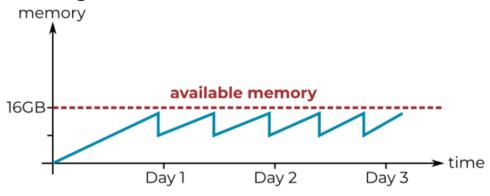
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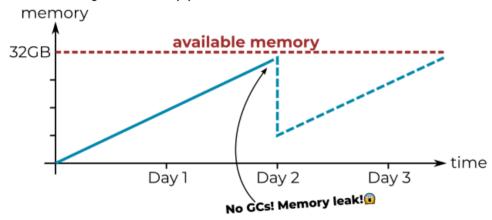
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4. Or we could solve "the memory leak" by reducing available RAM:

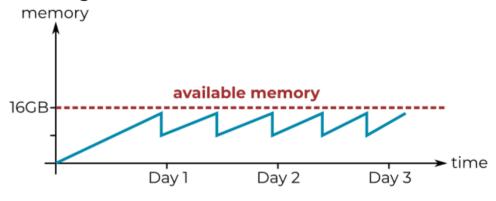


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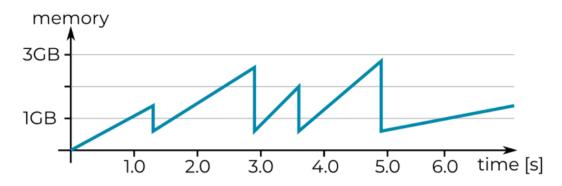
GC's aggressiveness may be its **implementation detail**, but we can have some **control** over it - we will return to that when understanding the GC a little more \mathfrak{S}

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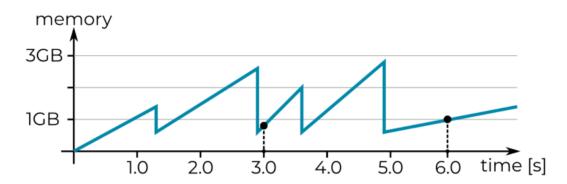
How to measure?

How to measure? We will return to that when understanding GC more. In general - it is "how much memory does this process use?" question.

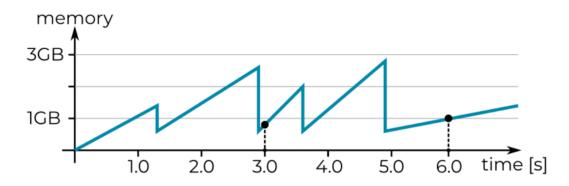
But, remember about one little fact. Memory usage change in time due to allocations & GC:



If we take a memory dumps at some random times:

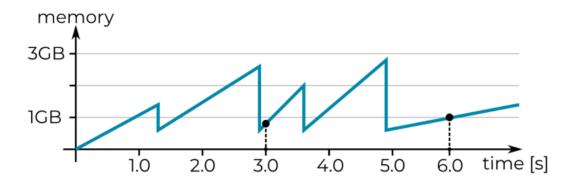


If we take a memory dumps at some random times:



.. we will see memory usage of around 1GB, not considering the dynamic nature of the GC!

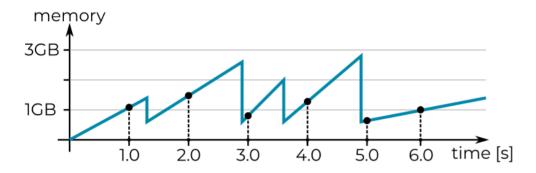
If we take a memory dumps at some random times:



.. we will see memory usage of around 1GB, not considering the dynamic nature of the GC! (and I see it surprisingly often to analyze memory dump without knowledge how it relates to the surrounding GCs)

If we measure on some regular, timely basis:

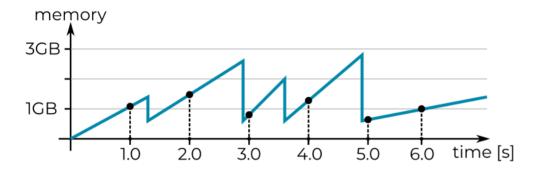
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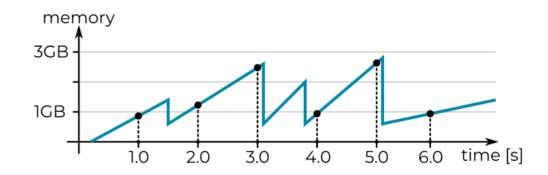
we will see 1, 1.5, 0.8, 1.2, 0.6, 1 GB memory usage

Memory overhead

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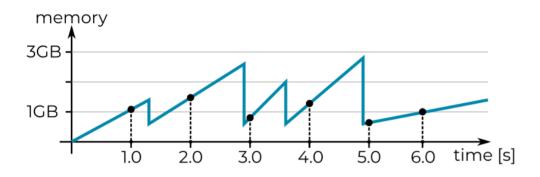
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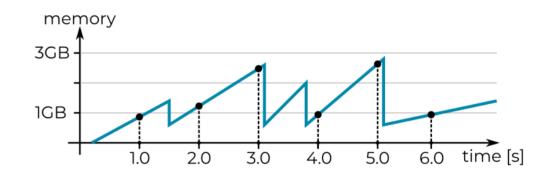


or 0.9, 1.2, **2.5**, 1.0, **2.6**, 1.0 GB. It really depends how "lucky" we are.

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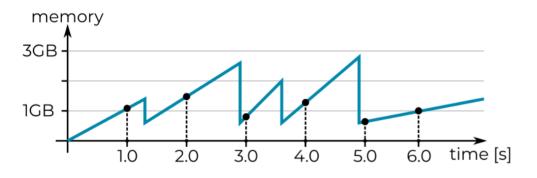
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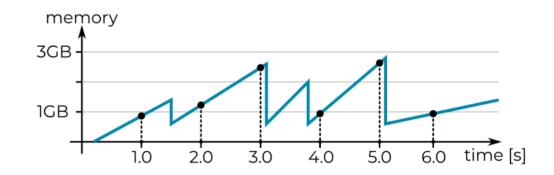
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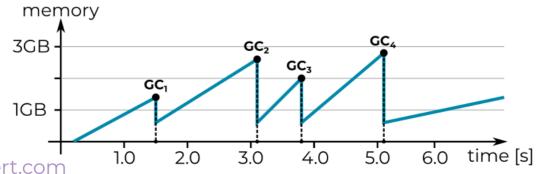




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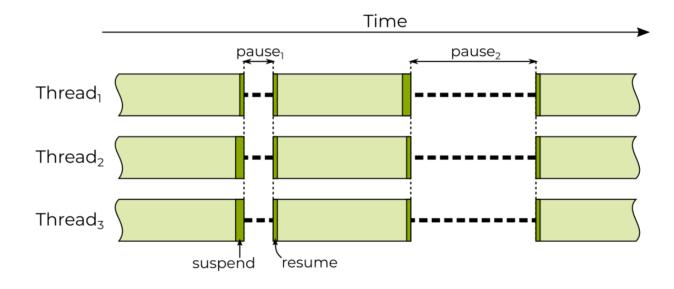
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- this is a typical *sampling problem* we may sample very often...
- or even better measure with respect to the GC to know exact values before and after:



12/37

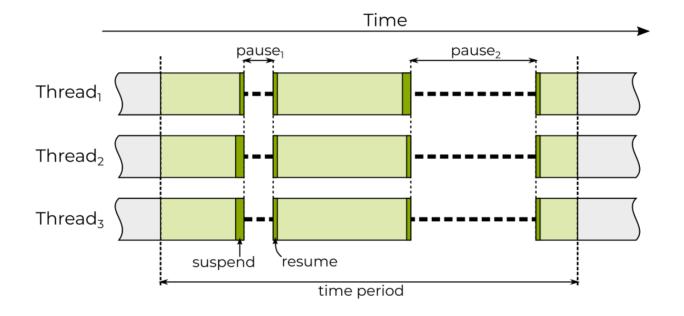
All GCs need to occasionally pause an application (threads) for shorter or longer amount of time:



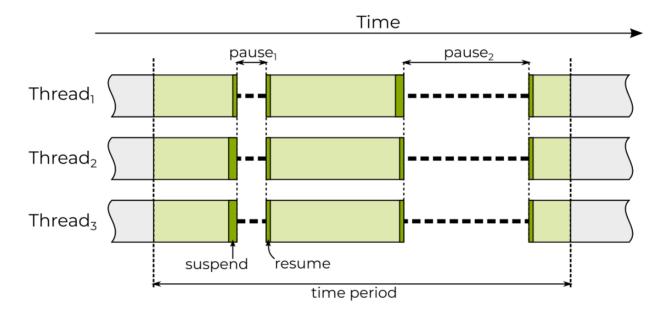
This is typically not-so-good:

- in interactive application it affects the user who may notice worse responsiveness
- in request processing application (like web) it affects some requests processing times

We can intuitively reflect it by relative pause time of the application:

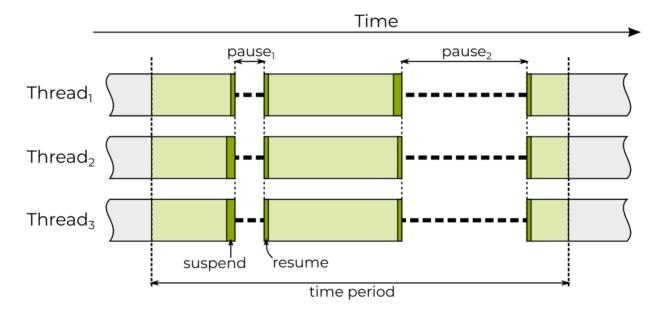


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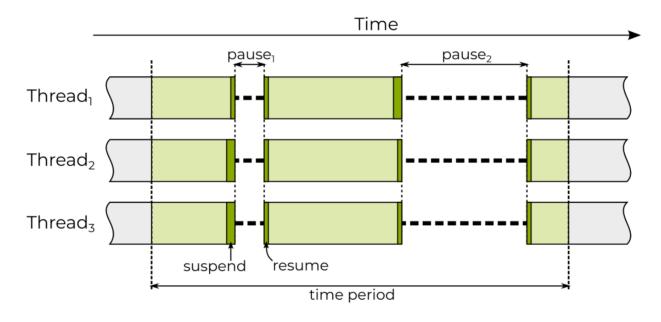
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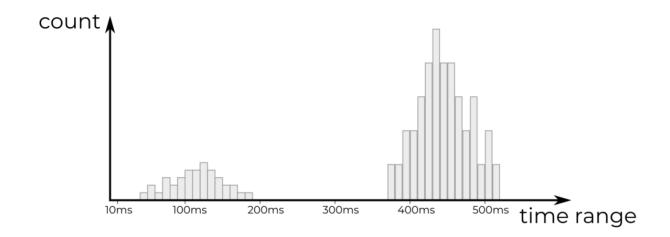
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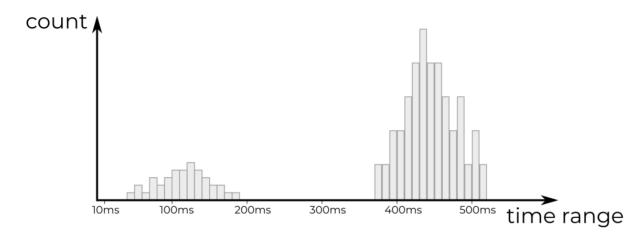
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"Usually a well behaved app has < 5% Pause time in GC while it's actively handling workloads."

An interesting point of view is to see **pauses histogram** - it may reveal a little bit more complex (and useful) image that a single % *Pause Time in GC*:



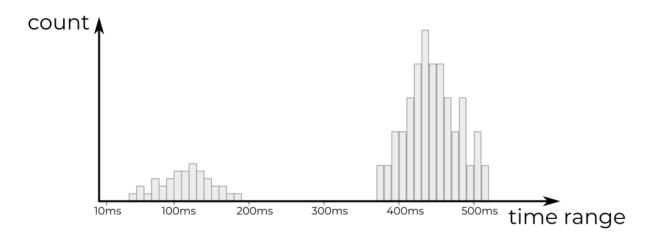
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- shorter pauses between 50-200 ms
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Obviously, we may be interested in reducing those longer ones. Especially since they affect so-called **tail latency**.

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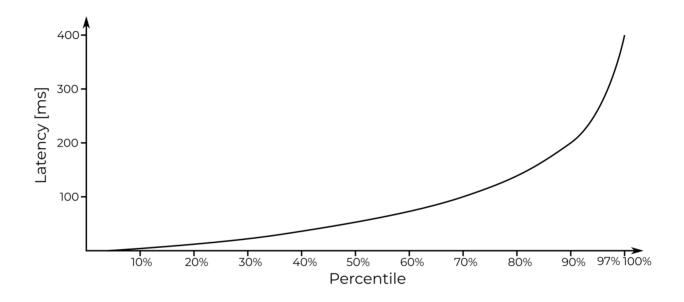
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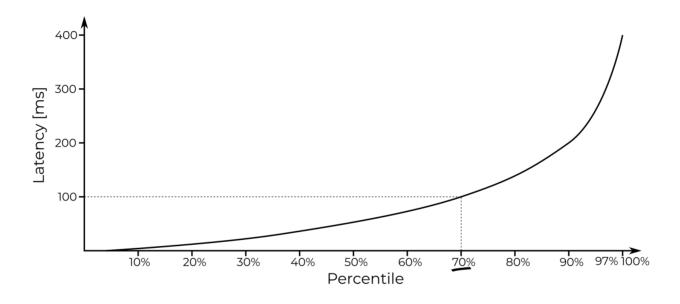
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- we may be interested in high 95-th (P95) or 99-th percentile (P99):
 - it is fine that 95% or 99% of customers are satisfied by fast response times
 - ∘ but... still there is this 5% or 1% which may be hit be the long latency (and leave with their 💸)

We can get a broad picture by visualizing latency in spectrum of percentiles:

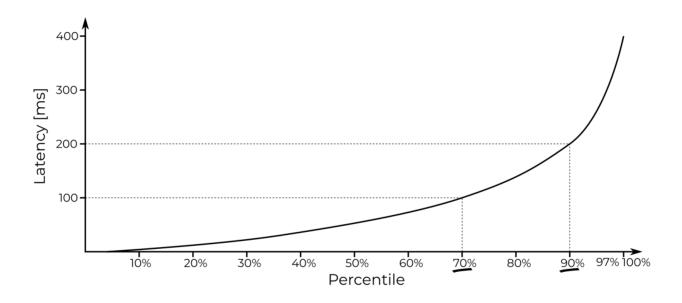


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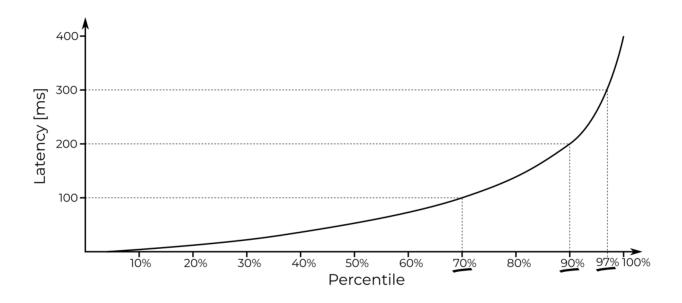
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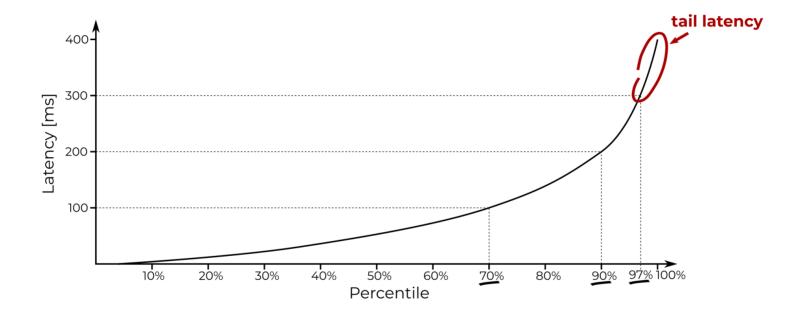


- **70-th percentile** 70% of responses are below 100 ms
- **90-th percentile** 90% of responses are below 200 ms (which leaves 10% people observing times slower than 200ms)

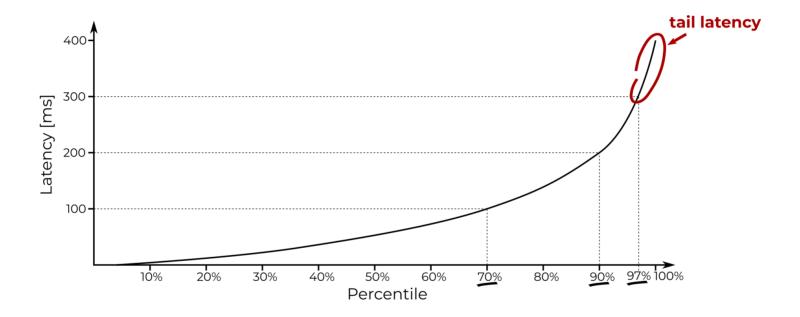
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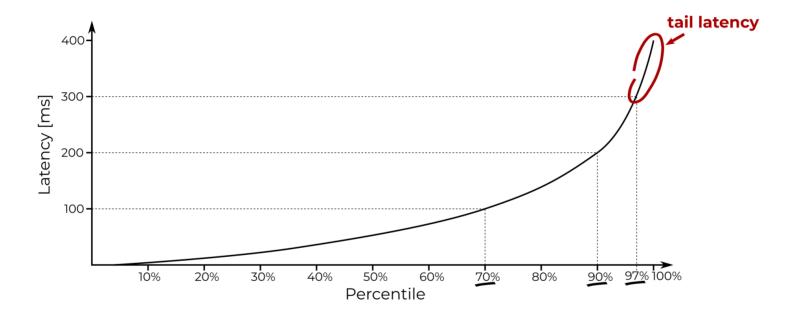
- **70-th percentile** 70% of responses are below 100 ms
- **90-th percentile** 90% of responses are below 200 ms (which leaves 10% people observing times slower than 200ms)
- **97-th percentile** 97% of responses are below 300 ms (which leaves 3% people observing times slower than 300ms)



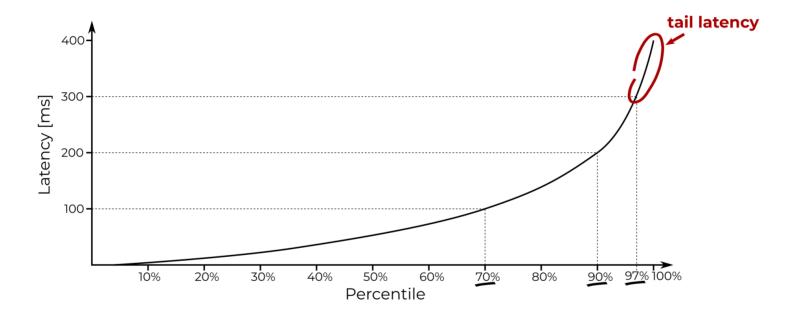
Tail latency - some high-percentile latency, observed by users rarely, but probably really severe:



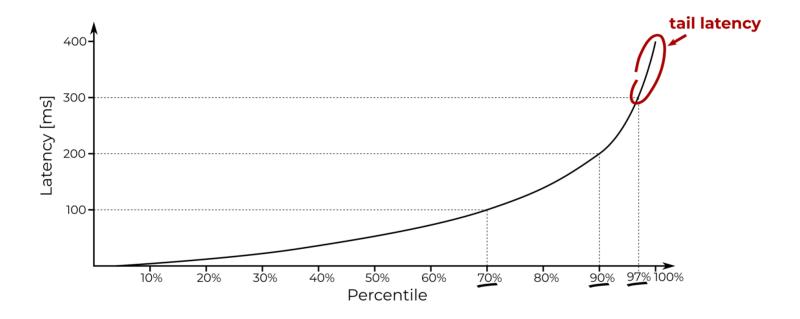
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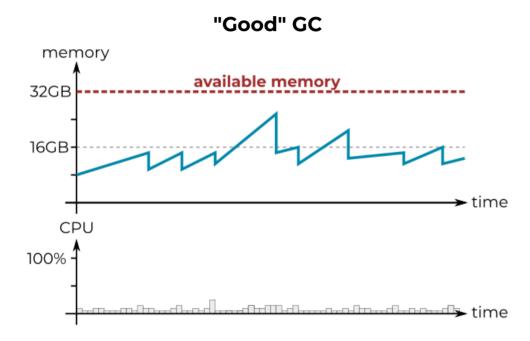
- optimizing tail latency is trying to reduce latency in some high percentile like P95 or P99
- for example, if you want to reduce P97 latency, look at responses <300 ms
 - we investigate reasons of responses slower than 300ms
 - ...and GC pauses may be one of them!

CPU overhead

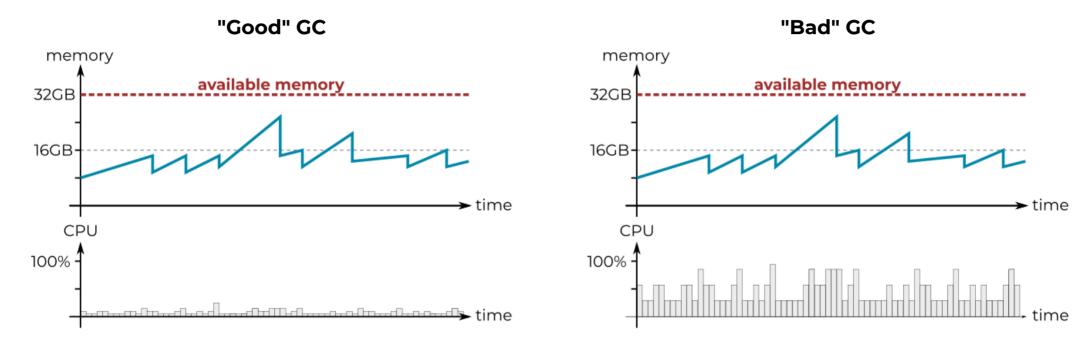
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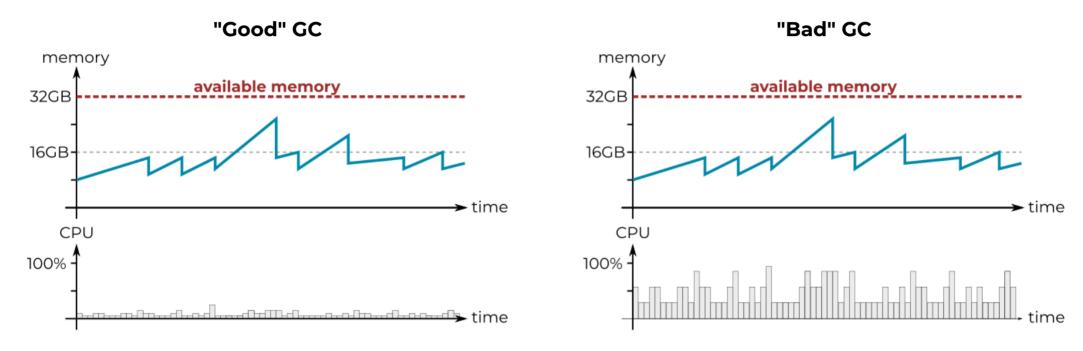
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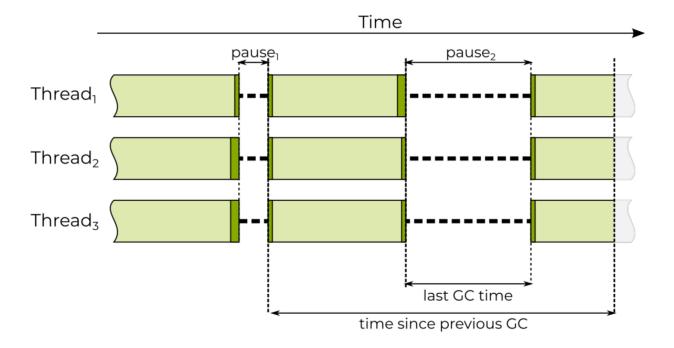


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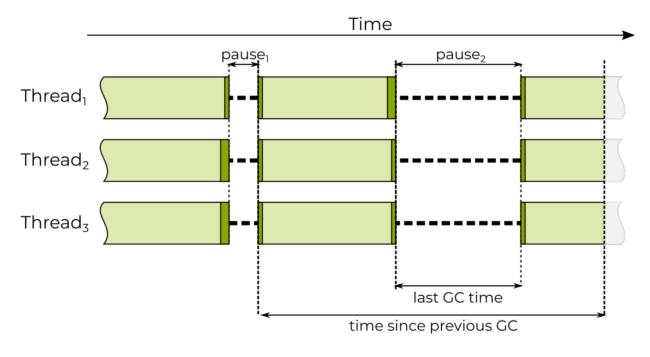


We could measure the whole machine CPU usage or our process CPU usage.

The simplest measurement is the relative time spent in the CC to the application time:

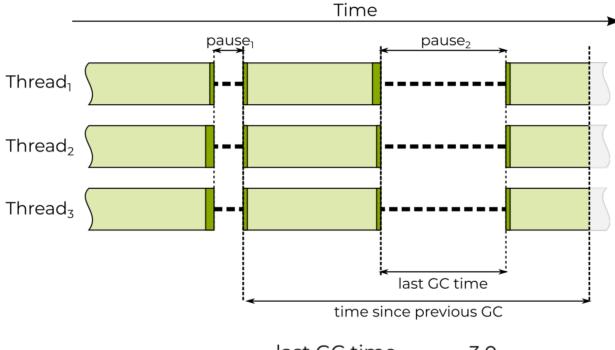


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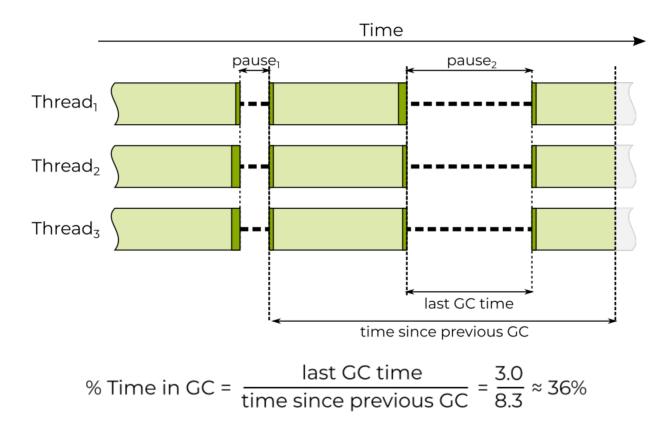
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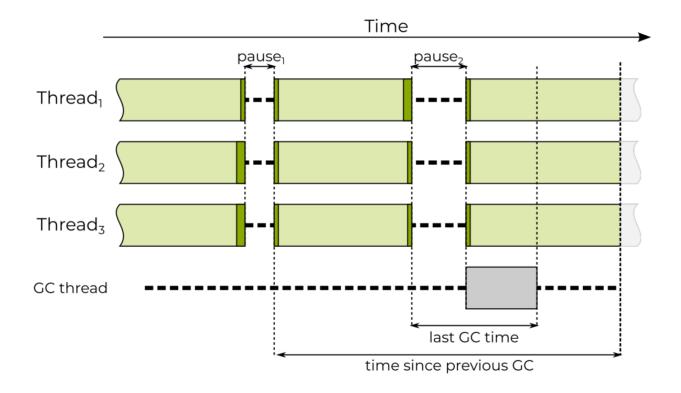
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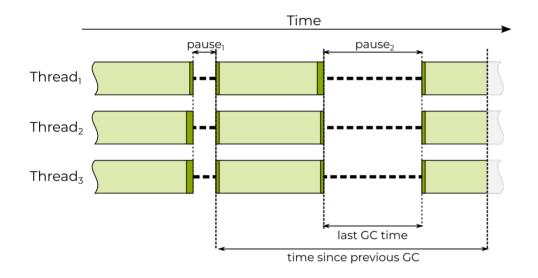


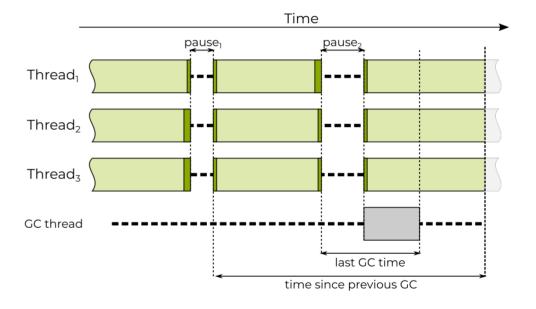
In case of GC work done only during pauses - it's the same as % Pause time in GC (as above).

If part of the GC work is done **concurrently**, it also counts:

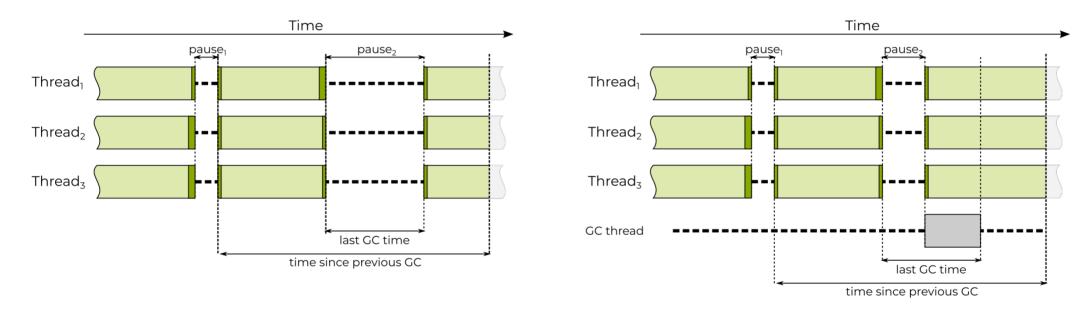


% Time in GC may be a little misleading - it produces the same values for concurrent and nonconcurrent case:



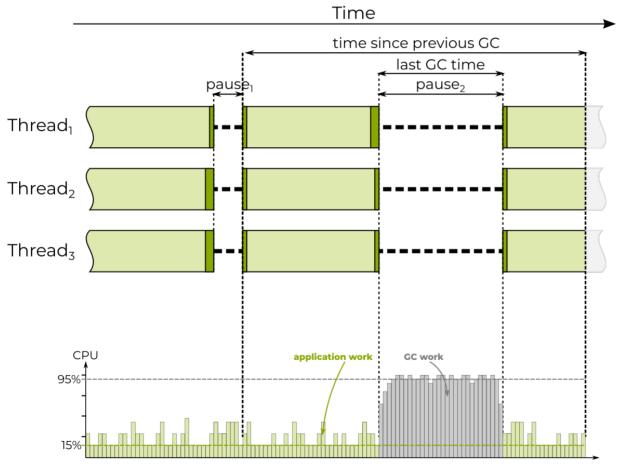


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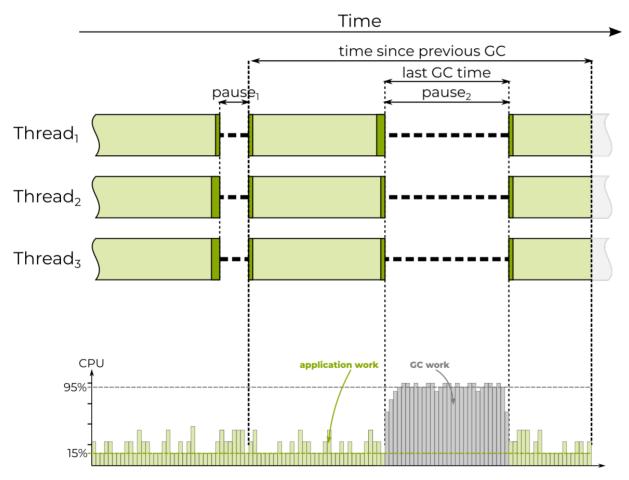
But we don't know which case consumes more CPU - it depends how CPU-intensive is the GC work during pause and concurrent phase.

% CPU time in GC is much more precise - instead of time, measure real CPU samples done by the CPU:



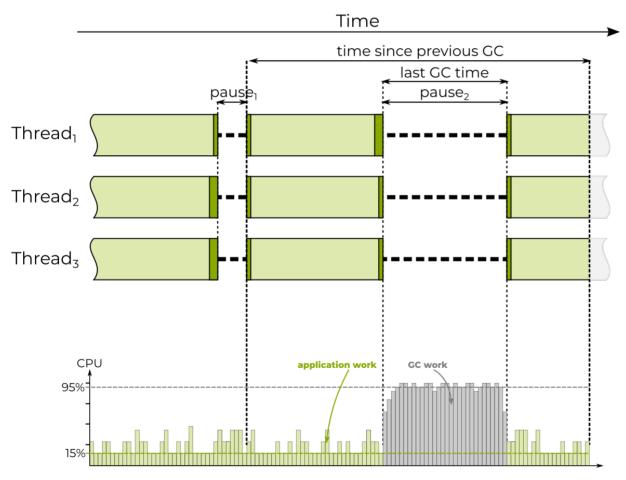
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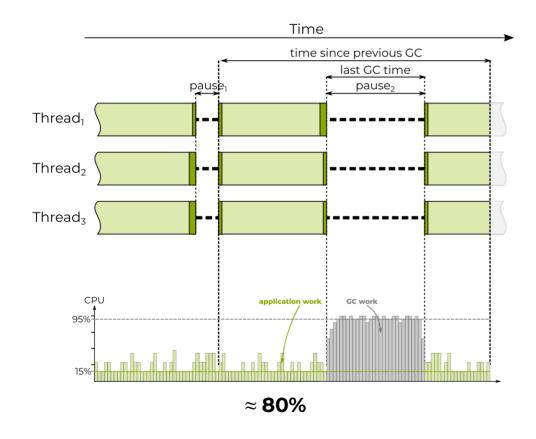


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- easy to calculate:

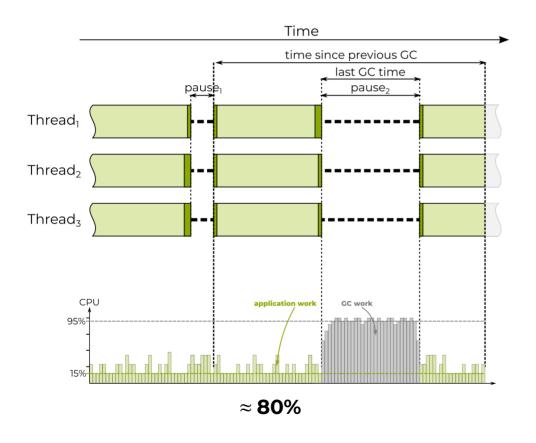
% CPU time in GC =
$$\frac{\text{CPU samples in GC}}{\text{total CPU samples}} \approx 80\%$$

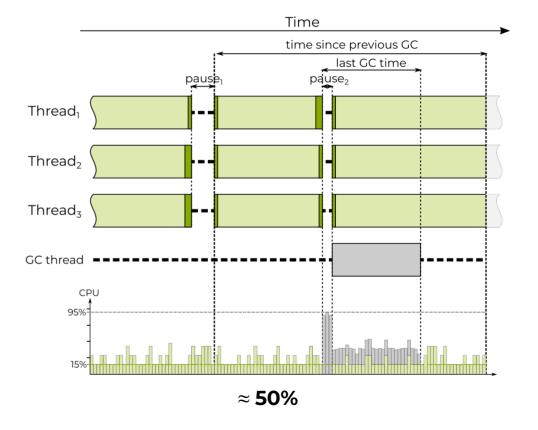
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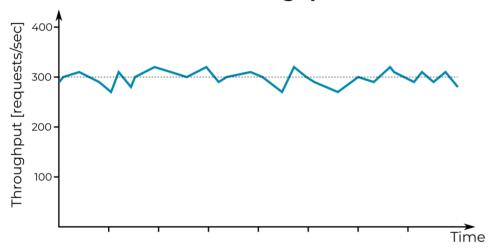
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- but is has its cost like resource (CPU) consumption etc.

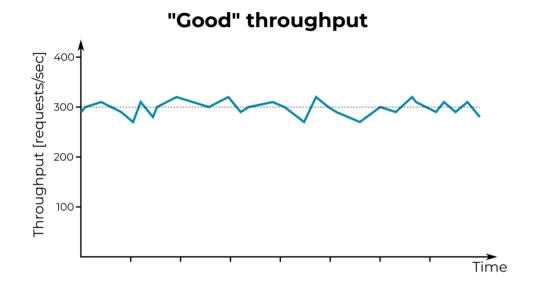
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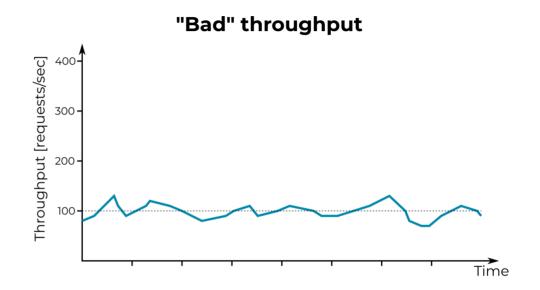
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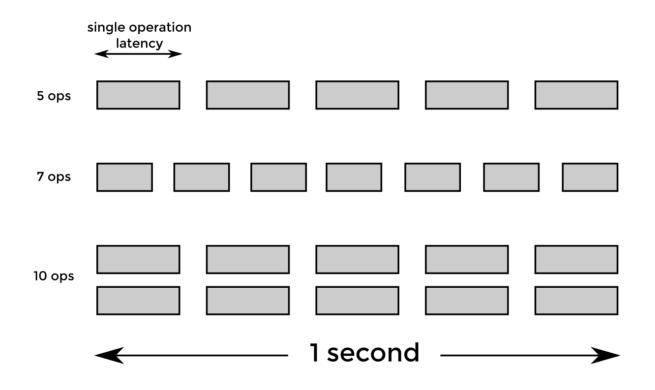
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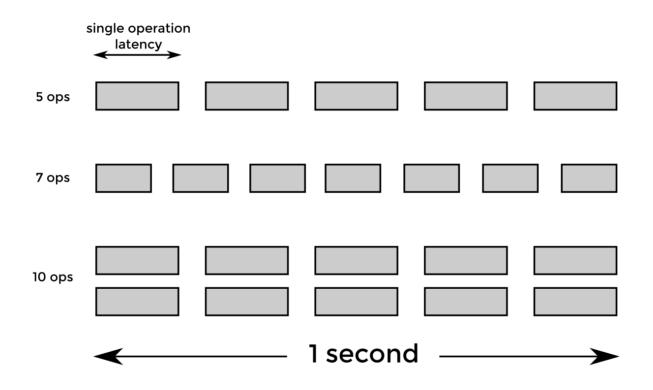
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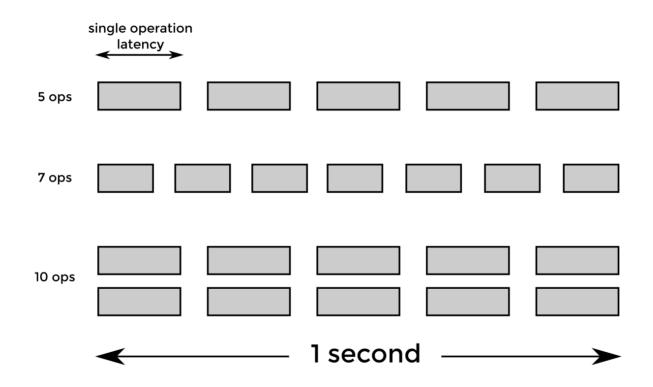
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- the problem is... there is no such a perfect GC
 - obviously we can optimize algorithms to make it more efficient
 - but in the end, pretty often we optimize one aspect as a trade off with the others





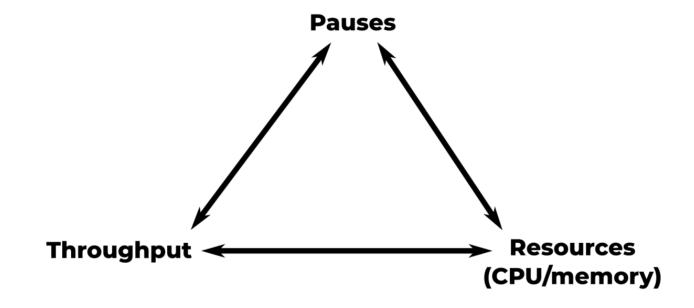
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- we can increase **throughput** by reducing **latency** (e.g. **GC pauses**)
- we can increase **throughput** by parallelizing work (e.g. consuming more CPU cores)

Garbage Collection Trilemma

The "impossible triangle" of the GC features:



Materials

• .NET Memory Performance Analysis