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Just **how mean**
can an average be?

*A visual inquiry on queue
dynamics and lead times
in supply chains.*



Cave canem (Beware of Dog) - Mosaic at the House of the Tragic Poet, Pompeii, 2nd century BC

The charts depicted are from the **Queue simulator** accompanying program.



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Challenging lead times

The aim of this inquiry is to clarify a key aspect of the **Core Conflict Area**^(*): **make visible** how **variation** affects queues and **impacts the performance** of a supply chain.

As queues are the largest contributors to lead times, under **what conditions** are stated lead times **relevant information**?

A magnifying glass with a wooden handle is positioned over a rectangular box. The box contains the text "ΔVisibility → ΔVariability" and "Core Conflict Area".

Δ Visibility \rightarrow Δ Variability
Core Conflict Area

(*) From *Demand Driven Performance*, D. Smith, Ch. Smith, 2014



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The Doctor's orders

A System of Profound Knowledge^(*)



Appreciation for a system (S)
Theory of knowledge (K)
Knowledge about variation (V)
Psychology (H)



(*) Dr. W. Edwards Deming (1900-1993) in *The New Economics* (1994).

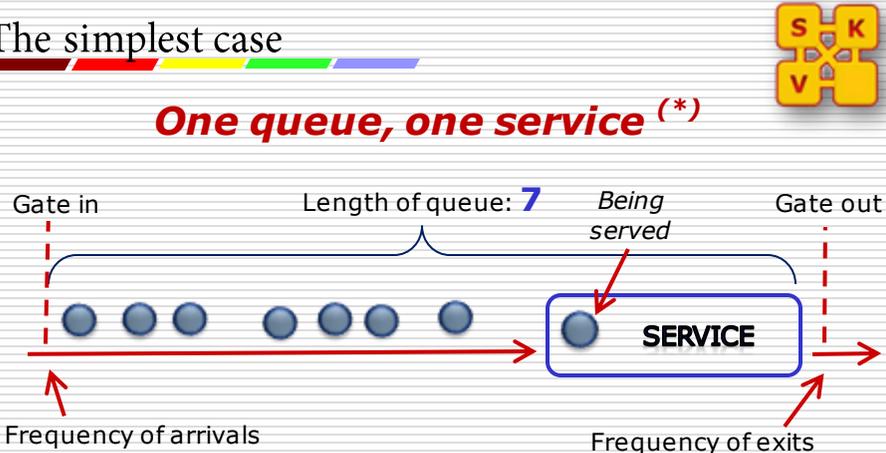
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The simplest case

One queue, one service^(*)




$$\text{Utilization} = \frac{\text{frequency of arrivals}}{\text{frequency of exits}}$$

(*) G/G/1, in Kendall's notation: both arrivals and service General distribution, 1 server, First Come First Served discipline.

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A primary question



If, at your **post office**, the operator **serves each client in one minute sharp**, how many **clients** can he/she serve **in 1 hour**?



60 clients, you say?

A dangerous relationship

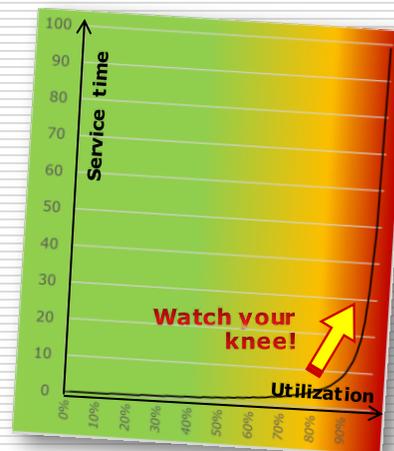


The **Little's Law** defines (*) the relationship between utilization (**U**) and the overall service time (**T**):

$$T \approx \frac{U}{1 - U}$$

The Law predicts the **long-term, steady-state average** length of the queue.

After a point, service time **increases sharply with utilization.**

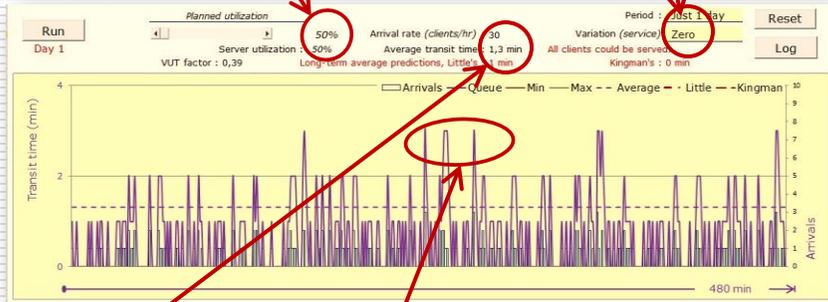


(*) Applies to M/M/1, First Come First Served queues

Let's check this out!



Using the **simulator**, we try first with **30 clients (utilization 50%)** and a **perfect clerk** who always takes **exactly 60 seconds** per client.



During the whole 8-hour opening, all clients are served in **1.3 minutes** in average.

Apart for a **few peaks of 3**, the line is always **very short and predictable**.

A tripping point

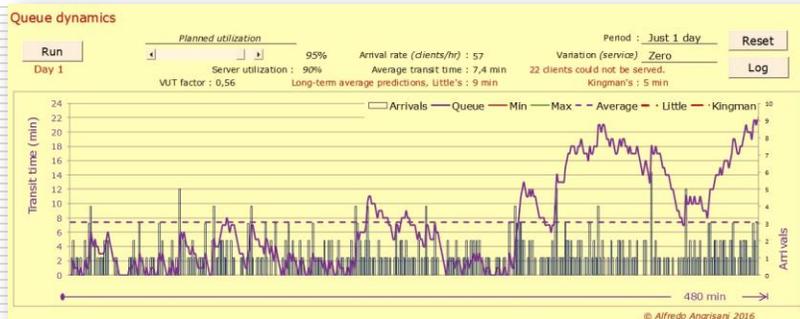


Utilization **85%**: several clients experience a wait two times the average (**6.3 minutes**), and one **cannot be served at all** during the opening hours (the clerk is absolutely reliable but strict!).

Instability gets worse and the **average doesn't seem to be a useful indicator any longer**.



Let's raise the stake!



Utilization **95%**: clients wait **7.4 minutes** in average (several of them more than **twice as long**) and **22 will not be served at all today**.

Average is definitely **misleading for most practical purposes!**

From Sir John Frank Charles Kingman:



The **VUT Equation** defines^(*) the combined effect of **Variation, Utilization** and **Time** for heavily-utilized services:



$$T_q \approx \left(\frac{c_a^2 + c_e^2}{2} \right) \left(\frac{u}{1-u} \right)$$

The **coefficients of variation, $CoV^{(**)}$** , of arrivals (c_a) and exits (c_e) have a **quadratic effect** on service time.

(*) Applies to G/G/1, First Come First Served queues

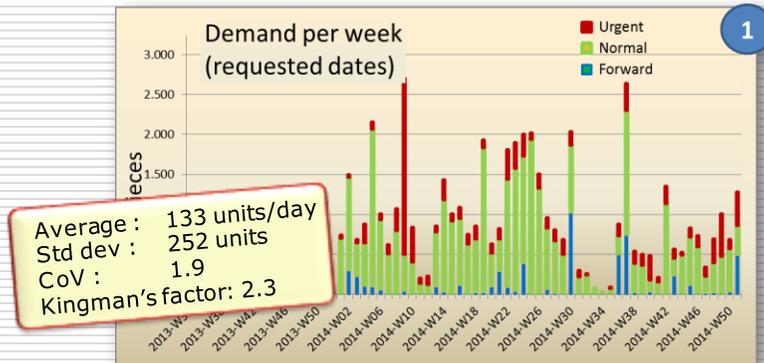
(**) For a given set of data, $CoV = \text{standard deviation} / \text{mean}$

The shape of demand



Mechanical assemblies: 35,000 groups,
286 PNs as sold worldwide to 400+ OEMs (*)

What the clients **wanted**:



(*) Example and graph, real data from the **DDMRP-DBR Simulator Demo - Gears**



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Waves you surf,
tsunamis you just
don't!



Some **shaking**
consequences



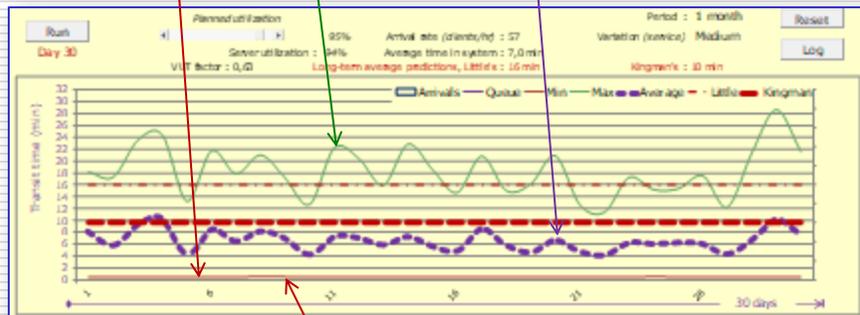
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Long-term **stability**, daily **rodeos**



The simulation run over a month shows the range of variation (**min** and **max**) and the **average** length of the queue **each day**:



Although the average transit time is 7 min, the queue goes as high as 28 and is **zero** -at some moment- **every day!**

Drawbacks



What do these **highs & lows** cause in **daily management**?

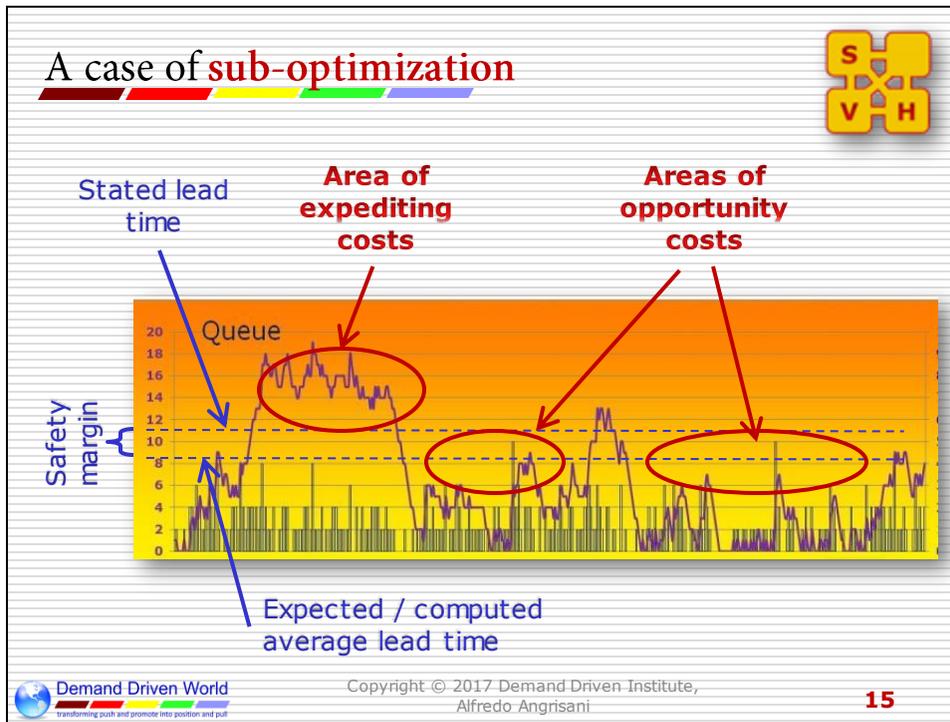


Highs:

- *Erratic priorities,*
- *Bad multitasking,*
- *Expediting, overtime costs*
- *Loss of service performance.*

Lows:

- *Potential work disruptions,*
- *Potential loss of market opportunities,*
- *Misjudgments about the real process capability.*



Lead times: **self-fulfilling prophecies?**

If you are **late**, you will:

- Do your best to recover the delay and thus incur in the **costs of expediting**,
- Accept a **loss in service** or in **product quality**,
- If delays are frequent, you may **increase the slack** without being aware of the consequences on the **flow of cash** and **materials** and on the **response to market**.

If you are **early**, you will:

- As the **Parkinson's Law** indicates, delay the work and, having used up the safety capacity, risk **to be late** if anything goes wrong,
- Don't exploit the advance, and incur in **opportunity costs**.
- If these delays are frequent, possibly deem that the lead time in use is **too long** and **reduce it!**

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Make a long queue short

A way to solution in
8 easy (*) recipes



(*) Well, maybe not always so easy!

A recipe in 8 easy steps – Part 1

- 1. Reduce the number** of critical resources in your flow
 - **Identify** the resources that exhibit capacity issues
 - **Optimize** the work of actual or potentially constrained resources
 - Consider **Lean initiatives** like **SMED**.
- 2. Feed** critically constrained resources from **controlled sources**
 - For this and the next points, follow the prescriptions of **DDMRP**
 - **Place buffers in front** of them and release the work **orderly**, at the **maximum rate allowed** by the resources themselves
- 3. Don't plan** for utilization **above 85%**
 - Don't go after **misleading efficiency KPIs!**
 - **Look ahead** for potential overloads



A recipe in 8 easy steps – Part 2

4. **Shorten** the lead times **by system design**
 - **Position buffers** so to **'slice' cumulative lead times** into shorter, independent segments.
5. **Stifle** the overall **process variation**
 - **By system design** using strategic buffers.
 - Consider **6 Sigma** initiatives.
6. **Reduce dependence** on lead times
 - **Shift the sequence** of control and appraisal **from compliance** with lead times (due dates) to the ability **to coordinate order priorities to the actual demand**
7. **Reduce variation** in demand
 - **Engineer the sales process** with **SPE**
 - **Enquire** about the **origin of peaks** and **mitigate** through the appropriate sales/supply chain **actions like VMI**.
 - Perform the **S&OP** process
8. **Adopt** flow-based, **Smart Metrics**
 - Use these outcomes to drive the **PDCA improvement cycle**