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Citation: Peeroo, K., Stankovic, V., Popov, P. & Weyde, T. (2024). Visualising the Impact of High-dimensional Configuration Parameters on the Performance of Data Distribution Service

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Visualising the Impact of High-dimensional Configuration **Parameters** on the **Performance** of Data Distribution Service

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Data Distribution Service (DDS) is a specification for middleware in real-time and mission-critical systems such as energy management systems and medical devices. Such systems have specific performance requirements to function properly and avoid catastrophe. However, the performance of the system can vary depending on the various parameters defined before or during runtime. Understanding how the different parameters impact performance helps DDS system designers set up distributed systems in critical domains. This work introduces a workflow for understanding how parameter values can affect system performance, as seen in their distribution form. It applies to DDS and any configurable system with multiple parameters that can take multiple values that may result in different system performance. This visualisation workflow is demonstrated on a dataset created by running tests on an actual DDS system. We argue that they help understand non-linear relationships between the parameters and DDS performance and help identify anomalous results that may need investigation.



their interdependencies on the performance of a DDS system?

Let's assume we are interested in how the stated parameters affect the latency of a DDS system.

1. Explore the global view using small multiples.

We start with the small multiples view seen at the top right. Parameters vary per row. Cumulative quantiles increase towards the right per **column**. Depending on the parameter value type, both line graphs and bar charts are used. Here, we look for irregularities in trends.

durability

Given that we are interested in latency, the worst-case latency can be seen in the last four columns. We would expect the latency to increase linearly as the number of publishers or subscribers increases. The top right cell (publisher count for the 99th percentile) shows otherwise.

2. Zoom into a specific quantile and generate median box plots.

We "zoom" into this column by generating box plots of the median latencies for all tests, as seen in the figure outlined in pink.

3. Filter out values for specific parameters.

Let's **isolate** all tests that use **10 publishers** to see what is happening. We **replot** the **box plots** as seen in the **figure** surrounded in blue.

4. Generate new median box plots.

The newly created blue figure shows a **new irregularity**: **25 subscribers** on top of 10 publishers have **higher latencies** than 50 subscribers.

5. Go back to step 3 (filter out values for specific parameters) until a conclusion is made.

Let's isolate all tests with 10 publishers and 25 subscribers and generate the box plots outlined in orange. We notice that 8 tests produced these values. That confirms that there were not a few anomalous results that skewed the data. The thing that pops out here is seen when we compare the blue figure with the orange figure. Focus on the use of multicast. The blue figure shows that using multicast (giving a value of 1) leads to lower latency. Meanwhile, the orange figure shows that using multicast leads to a higher latency. What is the difference between the two figures? The specific usage of 25 subscribers alongside 10 publishers.

6. Conclusion is made.

If a DDS system designer wants to design a system specifically, it is best not to use multicast to get more performant latency readings when using 10 publishers and 25 subscribers.

We introduce a global-to-local visualisation workflow that's only 5 steps:

- 1. Explore the global view using small multiples.
- 2. Zoom into a specific quantile and generate median box plots.
- 3. Filter out values for specific parameters.
- 4. Generate new median box plots.
- 5. Go back to step 3 until a conclusion is made.



- Creating a **tool** to **let users** interact in real-time.
- Deal with **limitations**:
 - Increasing the number of parameters or x-axis values makes the small multiples view visually overwhelming.
 - If tests with one parameter value produce really small latencies while tests with another parameter value produces really big latencies the box plots will be squashed out of proportion.
- Explore other options of multivariate visualisation that involve user interaction.

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