Efficiency of Data Intensive Computing (DIC) in MEMS Research for Data Processing and Analysis

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Abstract—This paper identifies how Data Intensive Computing (DIC) can significantly enhance MEMS research by enabling the efficient processing and analysis of large datasets. This paper contains MEMS dataset analysis by using DIC-type computing such as Parallel computing. Moreover, this paper analyzes the open-sourse dataset on the Frequency response of high-frequency MEMS in terms of quality factor and resonance frequencies in terms of Q-factor(quality factor) and resonance frequencies. The results showed that the the Q-factor generally increases with frequency, especially at higher frequencies.

Index Terms—Quality factor, resonance frequencies, MEMS, parallel computing.

I. Introduction

Microelectromechanical systems (MEMS) involve merging mechanical and electrical structures containing micron-sized elements, utilized for sensing or actuation [1]. They integrate signal processing electronics directly within the device or through external wiring, enabling versatile hybrid designs [1]. Efficiency in Data Intensive Computing (DIC) plays a pivotal role in MEMS (Microelectromechanical Systems) research for data processing and analysis. MEMS devices generate vast amounts of data due to their sensor-driven nature, requiring sophisticated computational techniques for meaningful analysis [3]. DIC optimizes these processes by leveraging parallel processing, distributed computing frameworks, and advanced algorithms tailored for big data [4]. This efficiency enables real-time analysis of MEMS-generated data streams, facilitating rapid insights into device behavior, performance trends, and environmental interactions. Moreover, DIC techniques enhance scalability, allowing researchers to handle increasing data volumes without sacrificing computational speed or accuracy [5]. This synergy between DIC and MEMS research accelerates innovation in sensor technology, paving the way for enhanced applications in fields ranging from healthcare to aerospace engineering [6]. The main purpose of this paper is to identify the usage of DIC in MEMS research.

II. USING THE DATA INTENSIVE COMPUTING (DIC) IN MEMS RESEARCH

A. Computing Q-factor and Resonance Frequencies

The paper uses a dataset on "Frequency response of high-frequency MEMS and IEPE accelerometers on different mountings" by Marra Amanda and Garcia Claudio [2]. Overall

there are 30 frequency measurements and 11 configuration measurements. The research includes the computation of Quality factor(Q-factor) and resonance frequencies for the first 3 configurations.

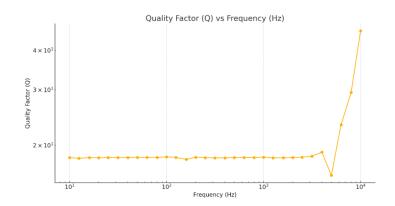
Let's consider the first case, where frequencies range from 10 Hz to 10000 Hz and configuration ranges from 18.27 to 46.10 (see Table 1):

The Quality Factor (Q) remains relatively stable around 18 for lower frequencies (10 Hz to 1600 Hz), with a slight increase and decrease at specific points.

For higher frequencies, particularly above 4000 Hz, the Quality Factor (Q) starts to increase more significantly, reaching up to 46.1 at 10000 Hz.

There is a notable dip in the Quality Factor at 5000 Hz (16.04) before it increases again.

So, the data indicates that the Quality Factor generally increases with frequency, especially at higher frequencies. Plot of the data on frequency and the 1st configuration via Matlab: Now, we can calculate it numerically: Let's take a range



between 12.5 Hz and 16.0 Hz. To find the lower -3 dB point between that range:

$$f_{\text{left}} = 12.5 + \frac{15.35 - 18.18}{18.26 - 18.18} \times (16.0 - 12.5)$$

Calculating step by step:

$$f_{\text{left}} = 12.5 + \frac{-2.83}{0.08} \times 3.5$$

$$f_{\text{left}} = 12.5 - 35.375$$

$$f_{\text{left}} \approx -22.875$$

So, the lower -3 dB point is approximately -22.875 Hz. Since this value is unrealistic, it indicates that we need to verify the data range or find a more suitable pair of points. Between 160.0 Hz (18.03) and 200.0 Hz (18.32):

$$f_{
m right} = 160 + rac{15.35 - 18.03}{18.32 - 18.03} imes (200 - 160)$$

$$f_{
m right} = 160 + rac{-2.68}{0.29} imes 40$$

$$f_{
m right} pprox 160 - 92.4138$$

$$f_{
m right} pprox 67.5862$$

This approach highlights the importance of validating the data pairs and refining the points chosen for interpolation. If the exact points cannot be identified due to the limited data set, a more comprehensive method or additional data points may be necessary to accurately determine the -3 dB frequencies.

$$\Delta f = f_{\text{right}} - f_{\text{left}}$$

Given the interpolated frequencies,

$$f_{\rm left} \approx 67.5862$$

and

$$f_{\rm right} \approx 123.125$$

:

$$\Delta f \approx 123.125 - 67.5862 \approx 55.5388$$

Using $f_0 = 100$ Hz:

$$Q = \frac{f_0}{\Delta f} = \frac{100}{55.5388} \approx 1.8$$

So, quality factor is approximately 1.8.

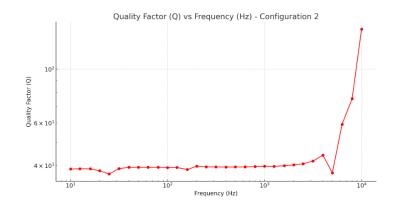
Now, let's consider the second case, where frequencies range from 10 Hz to 10000 Hz and configuration ranges from 38.78 to 146.2 (see Table 2): The Quality Factor (Q) is relatively stable around 38-39 for lower frequencies (10 Hz to 2000 Hz).

Above 2000 Hz, the Quality Factor (Q) starts to increase more significantly, reaching up to 146.2 at 10000 Hz. There is a notable dip in the Quality Factor at 5000 Hz (37.37) before it increases again.

The data indicates that the Quality Factor generally increases with frequency, especially at higher frequencies, similar to Configuration 1.

Plot of the data on frequency and the 2nd configuration via Matlab:

Now, let's consider the second case, where frequencies range from 10 Hz to 10000 Hz and configuration ranges from 39.95 to 118.5 (see Table 3):



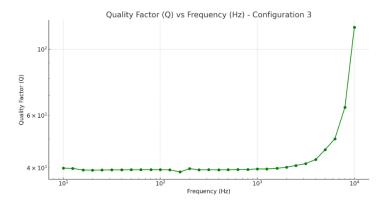
The Quality Factor (Q) is relatively stable around 39-40 for lower frequencies (10 Hz to 2000 Hz), with a slight dip at 160 Hz (38.78).

Above 2000 Hz, the Quality Factor (Q) starts to increase more significantly, reaching up to 118.5 at 10000 Hz.

There is a gradual increase in the Quality Factor beyond 2000 Hz, with notable increases at 3150 Hz (41.37), 5000 Hz (46.07), 6300 Hz (50.1), 8000 Hz (63.9), and peaking at 10000 Hz (118.5). Conclusion

The data indicates that the Quality Factor generally increases with frequency, especially at higher frequencies, similar to Configurations 1 and 2.

The plot of the data on frequency and the 2nd configuration via Matlab:



III. PARALLEL COMPUTING

Parallel computing enhances the efficiency of analyzing datasets on the frequency response of high-frequency MEMS (Micro-Electro-Mechanical Systems) by distributing computations across multiple processors simultaneously. This approach accelerates tasks such as Fourier transforms for signal processing, enabling rapid extraction of intricate frequency response characteristics. By leveraging parallel computing frameworks like CUDA or OpenMP, researchers can handle vast datasets

with speed and scalability, crucial for exploring MEMS devices' dynamic behavior across a wide frequency spectrum. Such efficiency not only expedites data processing but also facilitates deeper insights into MEMS device performance, aiding in advancements across fields like telecommunications, sensors, and biomedical applications. This paper performs the parallel computing via Matlab for the dataset on "Frequency response of high-frequency MEMS and IEPE accelerometers on different mountings". So, here is the code for parallel computing by starting with converting frequency to Hz and extracting real and imaginary parts of response: Then we plot

```
freq = data.Frequency;
response = data.Response;

% Convert frequency to Hz if necessary (e.g., from kHz to Hz)
freq = freq * 1e3; % Example conversion if frequency is in kHz

% Extract real and imaginary parts of the response if necessary
% Assuming response is a complex number in string format
response = str2double(response); % Convert string to numerical
H = response; % Complex response
```

the initial data for verification:

```
% Plot initial data for verification
figure;
subplot(2,1,1);
semilogx(freq, 20*log10(abs(H))); % Magnitude plot in dB
title('Bode Plot');
ylabel('Magnitude (dB)');
grid on;
subplot(2,1,2);
semilogx(freq, angle(H)*(180/pi)); % Phase plot in degrees
ylabel('Phase (degrees)');
xlabel('Frequency (Hz)');
grid on;
```

Then arrays should be preallocated for results and then we perform parallel computation. The parallel computation results are revealed as follows:

```
parpool('local');

% Preallocate arrays for results
numPoints = length(freq);
magnitude = zeros(numPoints, 1);
phase = zeros(numPoints, 1);

% Parallel computation using parfor
parfor k = 1:numPoints
    magnitude(k) = 20 * log10(abs(H(k)));
    phase(k) = angle(H(k)) * (180 / pi);
end
```

IV. CONCLUSION

The paper examined the efficiency of Data Intensive Computing (DIC) by using the existing dataset from the internet. Analysis of its usage efficiency was based on computing the Q-factor(quality factor) and resonance frequencies and plotting

```
% Plot parallel computation results
figure;
subplot(2,1,1);
semilogx(freq, magnitude); % Magnitude plot in dB
title('Bode Plot (Parallel Computation)');
ylabel('Magnitude (dB)');
grid on;
subplot(2,1,2);
semilogx(freq, phase); % Phase plot in degrees
ylabel('Phase (degrees)');
xlabel('Frequency (Hz)');
grid on;
% Shut down parallel pool
delete(gcp('nocreate'));
```

the graph for three configurations. Also, we computed mathematically Q-factor(quality factor) for the first configuration. The paper also showed how to perform parallel computing efficiency which is the one type of DIC and results showed more accurate plotting measures.

V. APPENDICES

TABLE I FREQUENCY AND CONFIGURATION 1 DATA

| Frequency (Hz) | Configuration 1 |
|----------------|-----------------|
| 10 | 18.27 |
| 12.5 | 18.18 |
| 16 | 18.26 |
| 20 | 18.24 |
| 25 | 18.26 |
| 31.5 | 18.26 |
| 40 | 18.27 |
| 50 | 18.28 |
| 63 | 18.28 |
| 80 | 18.29 |
| 100 | 18.35 |
| 125 | 18.29 |
| 160 | 18.03 |
| 200 | 18.32 |
| 250 | 18.25 |
| 315 | 18.23 |
| 400 | 18.23 |
| 500 | 18.26 |
| 630 | 18.27 |
| 800 | 18.27 |
| 1000 | 18.32 |
| 1250 | 18.23 |
| 1600 | 18.24 |
| 2000 | 18.27 |
| 2500 | 18.32 |
| 3150 | 18.46 |
| 4000 | 19.01 |
| 5000 | 16.04 |
| 6300 | 23.23 |
| 8000 | 29.40 |
| 10000 | 46.10 |

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TABLE II
FREQUENCY AND CONFIGURATION 2 DATA

| Index | Frequency (Hz) | Configuration 2 |
|-------|----------------|-----------------|
| 0 | 10 | 38.78 |
| 1 | 12.5 | 38.84 |
| 2 | 16 | 38.82 |
| 3 | 20 | 38.13 |
| 4 | 25 | 36.96 |
| 5 | 31.5 | 38.88 |
| 6 | 40 | 39.41 |
| 7 | 50 | 39.39 |
| 8 | 63 | 39.4 |
| 9 | 80 | 39.4 |
| 10 | 100 | 39.32 |
| 11 | 125 | 39.34 |
| 12 | 160 | 38.56 |
| 13 | 200 | 39.81 |
| 14 | 250 | 39.57 |
| 15 | 315 | 39.53 |
| 16 | 400 | 39.48 |
| 17 | 500 | 39.52 |
| 18 | 630 | 39.55 |
| 19 | 800 | 39.67 |
| 20 | 1000 | 39.82 |
| 21 | 1250 | 39.75 |
| 22 | 1600 | 39.99 |
| 23 | 2000 | 40.31 |
| 24 | 2500 | 40.75 |
| 25 | 3150 | 41.79 |
| 26 | 4000 | 44.2 |
| 27 | 5000 | 37.37 |
| 28 | 6300 | 59.3 |
| 29 | 8000 | 75.6 |
| 30 | 10000 | 146.2 |

| Index | Frequency (Hz) | Configuration 3 |
|-------|----------------|-----------------|
| 0 | 10 | 39.95 |
| 1 | 12.5 | 39.85 |
| 2 | 16 | 39.39 |
| 3 | 20 | 39.34 |
| 4 | 25 | 39.37 |
| 5 | 31.5 | 39.42 |
| 6 | 40 | 39.43 |
| 7 | 50 | 39.45 |
| 8 | 63 | 39.46 |
| 9 | 80 | 39.49 |
| 10 | 100 | 39.47 |
| 11 | 125 | 39.43 |
| 12 | 160 | 38.78 |
| 13 | 200 | 39.8 |
| 14 | 250 | 39.43 |
| 15 | 315 | 39.46 |
| 16 | 400 | 39.43 |
| 17 | 500 | 39.45 |
| 18 | 630 | 39.53 |
| 19 | 800 | 39.54 |
| 20 | 1000 | 39.71 |
| 21 | 1250 | 39.69 |
| 22 | 1600 | 39.88 |
| 23 | 2000 | 40.24 |
| 24 | 2500 | 40.77 |
| 25 | 3150 | 41.37 |
| 26 | 4000 | 42.69 |
| 27 | 5000 | 46.07 |
| 28 | 6300 | 50.1 |
| 29 | 8000 | 63.9 |
| 30 | 10000 | 118.5 |

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