

# COL851 Course Project

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# Contents

<b>1</b>	<b>Introduction</b>	<b>3</b>
1.1	Student details . . . . .	3
1.2	Project Objectives . . . . .	3
<b>2</b>	<b>Methodology</b>	<b>3</b>
2.1	Data Description and Pre-processing . . . . .	3
2.2	Chronos Model and Constraints . . . . .	3
<b>3</b>	<b>Results</b>	<b>4</b>
3.1	Performance Summary on Gurgaon (City A) . . . . .	4
3.2	Performance Summary on Patna (City B) . . . . .	7
<b>4</b>	<b>Discussion</b>	<b>10</b>
4.1	Optimal Configuration: Best Average RMSE for Both Cities . . . . .	10
4.2	Comparison between Cities . . . . .	10
<b>5</b>	<b>Performance and Resource Analysis</b>	<b>10</b>
5.1	Measurement Setup . . . . .	10
5.2	Observed Performance Trends . . . . .	11
5.3	Quantitative Highlights . . . . .	11
5.4	Visualizations . . . . .	11

# 1 Introduction

## 1.1 Student details

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## 1.2 Project Objectives

The primary goals of this project are:

1. To deploy and evaluate the zero-shot performance of the Chronos LLM for PM forecasting on datasets from \*\*Gurgaon (City A)\*\* and \*\*Patna (City B)\*\*.
2. To perform a parametric study on the effect of \*\*context length\*\* (2, 4, 8, 10, 14 days) and \*\*forecasting horizon\*\* (4, 8, 12, 24, 48 hours) on accuracy (RMSE).
3. To benchmark different feasible \*\*pre-trained model variants\*\* (`chronos-t5-mini`, `chronos-t5-small`, `chronos-bolt-small`) while adhering to the constraint of \*\*laptop CPU\*\* computation.
4. To identify the overall best-performing configuration (model, context, and horizon) based on the combined \*\*Average RMSE\*\* for both cities.

# 2 Methodology

## 2.1 Data Description and Pre-processing

The time series data comprises PM concentration records for Gurgaon (City A) and Patna (City B). For this experiment, only two columns were utilized: \*\*Column 1 (Time Index)\*\* and \*\*Column 5 (PM concentration)\*\*. The frequency of the data is assumed to be hourly. The data was partitioned into historical context windows and future prediction windows for rolling-window evaluation, using the \*\*Root Mean Squared Error (RMSE)\*\* as the accuracy metric.

## 2.2 Chronos Model and Constraints

The **Amazon Chronos** model was used in its \*\*Zero-Shot\*\* mode. Due to the \*\*laptop CPU constraint\*\*, three model variants were tested to identify the best trade-off between size and performance:

- `amazon/chronos-t5-mini`
- `amazon/chronos-t5-small`
- `amazon/chronos-bolt-small`

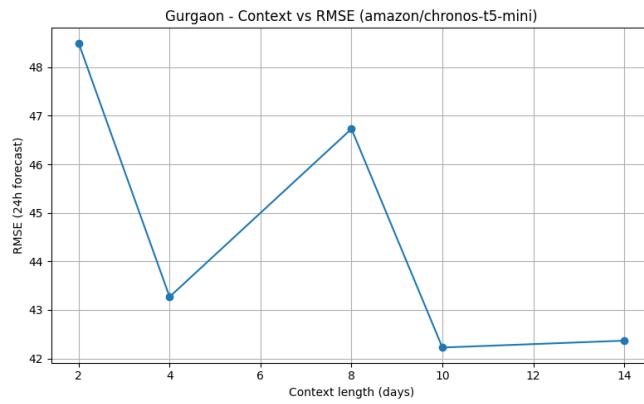
### 3 Results

#### 3.1 Performance Summary on Gurgaon (City A)

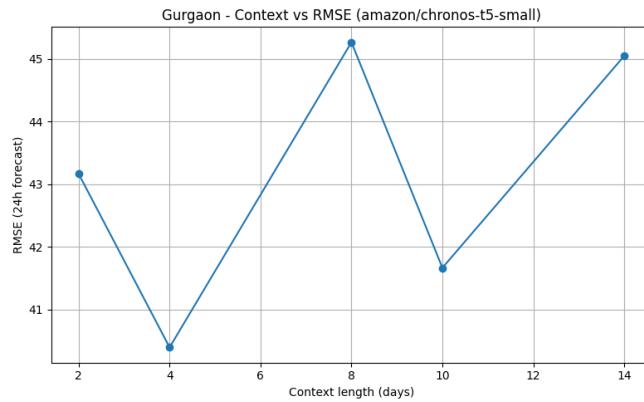
For Gurgaon (City A), the overall lowest RMSE was achieved at the shortest horizon by the `chronos-bolt-small` model. The feasibility check allowed for testing three variants: `chronos-t5-mini`, `chronos-t5-small`, and `chronos-bolt-small`.

Table 1: Summary of Best Zero-Shot Performance on Gurgaon Data (City A)

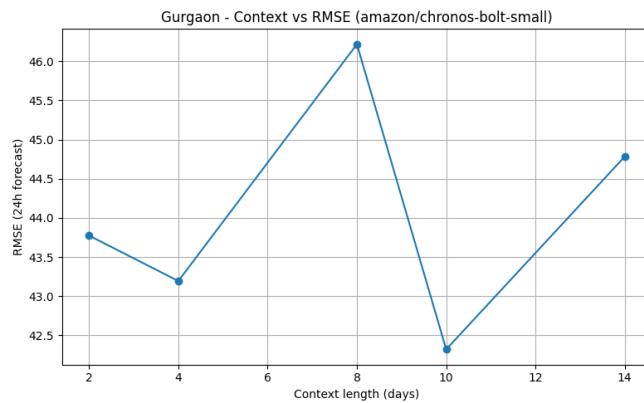
Model	Best Ctx (days)	Min RMSE (24h)	Best Hz (hrs)	Min RMSE (10d Ctx)
t5-mini	10	42.23	4	26.42
t5-small	4	<b>40.40</b>	4	26.34
bolt-small	10	42.32	4	<b>25.74</b>



(a) chronos-t5-mini Context

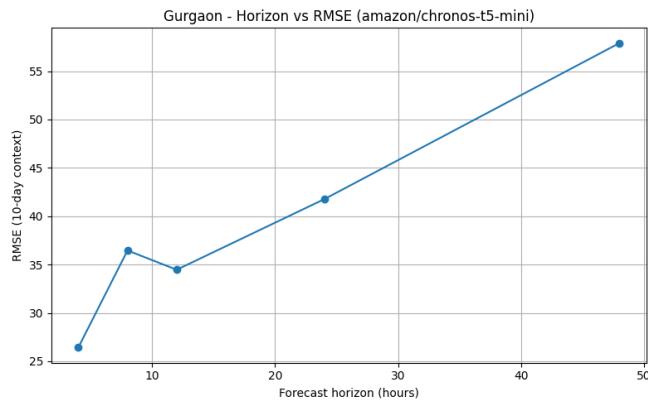


(b) chronos-t5-small Context

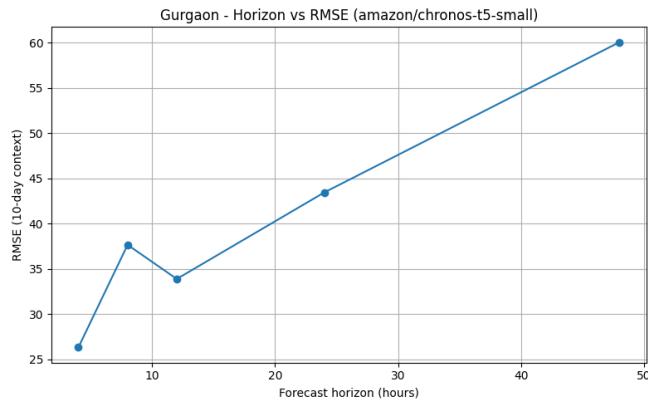


(c) chronos-bolt-small Context

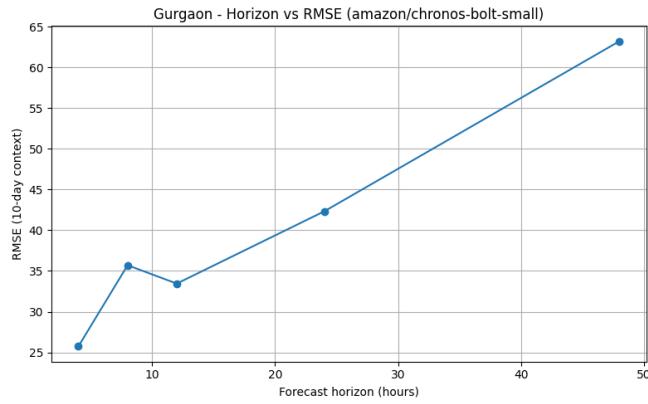
Figure 1: Average RMSE for PM forecasting (24h horizon) with varying context lengths (Gurgaon).



(a) chronos-t5-mini Horizon



(b) chronos-t5-small Horizon



(c) chronos-bolt-small Horizon

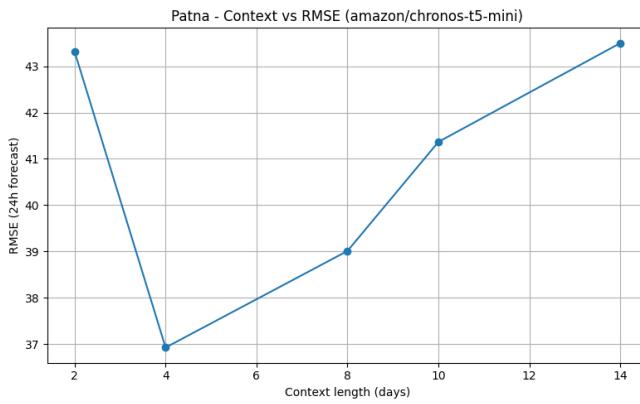
Figure 2: Average RMSE for PM forecasting (10-day context) with varying forecasting horizons (Gurgaon).

### 3.2 Performance Summary on Patna (City B)

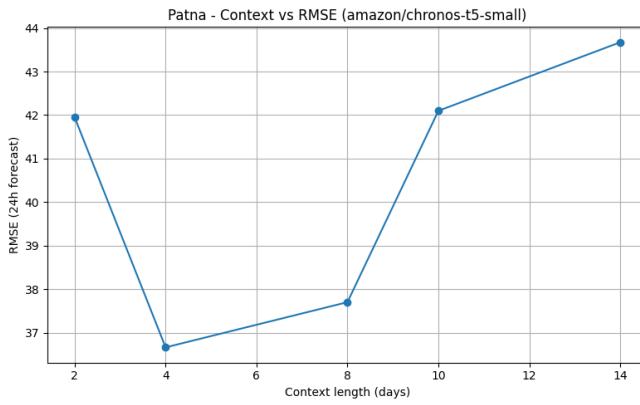
For Patna (City B), the `chronos-bolt-small` model also delivered the lowest RMSE, achieving slightly better overall performance than in Gurgaon.

Table 2: Summary of Best Zero-Shot Performance on Patna Data (City B)

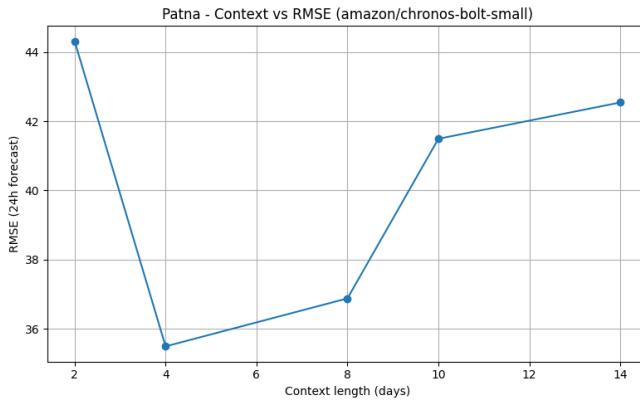
Model	Best Ctx (days)	Min RMSE (24h)	Best Hz (hrs)	Min RMSE (10d Ctx)
t5-mini	4	36.92	4	25.78
t5-small	4	36.67	4	25.09
bolt-small	4	<b>35.49</b>	4	<b>24.88</b>



(a) chronos-t5-mini Context

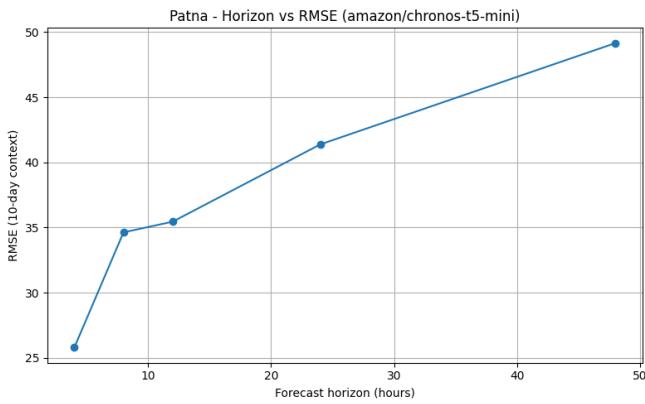


(b) chronos-t5-small Context

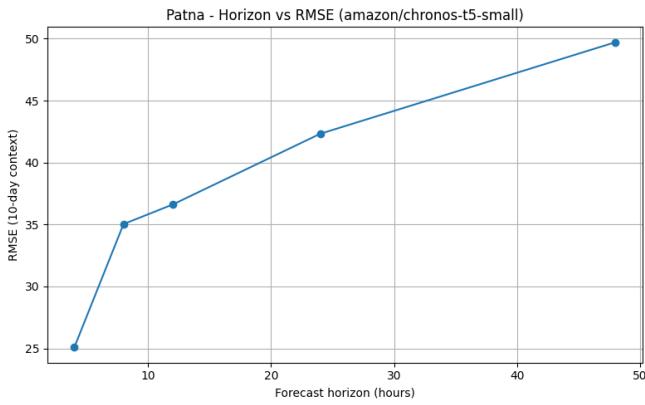


(c) chronos-bolt-small Context

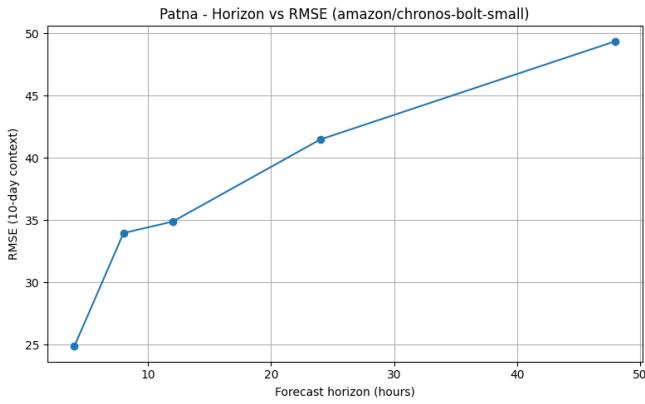
Figure 3: Average RMSE for PM forecasting (24h horizon) with varying context lengths (Patna).



(a) **chronos-t5-mini** Horizon



(b) **chronos-t5-small** Horizon



(c) **chronos-bolt-small** Horizon

Figure 4: Average RMSE for PM forecasting (10-day context) with varying forecasting horizons (Patna).

## 4 Discussion

### 4.1 Optimal Configuration: Best Average RMSE for Both Cities

The comprehensive analysis across both cities reveals a clear superior zero-shot configuration:

- **Best Model Variant:** `**amazon/chronos-bolt-small**` (Min RMSE  $\approx 24.88$ )
- **Best Forecasting Horizon:** `**4 hours**`
- **Best Context Length:** `**4 days**`

The `**chronos-bolt-small**` model consistently delivered the lowest RMSE in the most predictable scenario (4-hour horizon) for both cities, slightly outperforming the T5-based variants. The optimal context length, which minimizes RMSE for the challenging 24-hour horizon, was `**4 days**` for Patna, and either 4 or 10 days for Gurgaon depending on the model. This suggests that for highly non-linear data like PM, a relatively short context (around 4 days) is often sufficient to capture relevant daily/short-term patterns without adding unnecessary noise or computational load.

### 4.2 Comparison between Cities

- **Overall Accuracy:** Patna (City B) achieved a slightly lower overall minimum RMSE ( $\approx 24.88$ ) compared to Gurgaon (City A,  $\approx 25.74$ ). This indicates that the PM time series in Patna may exhibit slightly clearer or more stable patterns that the pre-trained Chronos model can identify in a zero-shot setting.
- **Context Length Preference:** Patna consistently favored a `**4-day**` context across all models for the 24-hour prediction task, while Gurgaon’s best context was either 4 or 10 days. This implies that Patna’s relevant forecasting signals may be concentrated in the immediate past, whereas Gurgaon’s patterns might occasionally benefit from capturing weekly or 10-day cycles.
- **Horizon Sensitivity:** Both cities demonstrated high sensitivity to the forecasting horizon. RMSE values increased steeply as the horizon moved from 4 hours to 48 hours, highlighting the inherent difficulty of zero-shot, univariate forecasting for volatile air quality data in the absence of co-variates.

## 5 Performance and Resource Analysis

### 5.1 Measurement Setup

Performance metrics were collected using `psutil` and Prometheus during zero-shot Chronos forecasts on CPU. Each configuration measured:

- Average and 95th-percentile inference latency
- Throughput (samples/sec)
- CPU utilization (%)

- Memory utilization (%)

Experiments varied context length (2–14 days) and horizon (4–48 h) across three models.

## 5.2 Observed Performance Trends

- **Latency:** Increased with longer context and horizon. For Gurgaon (`t5-mini`), average latency rose from 0.35 s (2 days) to 0.60 s (14 days).
- **Throughput:** Inversely related to latency; `chronos-bolt-small` reached  $\approx 37\text{--}42$  samples/s.
- **CPU Utilization:** Averaged 55–65 % across all settings, peaking near 76 % for longest contexts.

## 5.3 Quantitative Highlights

Table 3: Performance Summary (10-day context, 24-hour horizon, CPU)

Model	City	Latency (s)	Throughput	CPU (%)	MEM (%)	RMSE
<code>t5-mini</code>	Gurgaon	0.50	1.98	56.7	91.2	115.75
<code>t5-small</code>	Gurgaon	0.79	1.26	57.5	92.6	116.88
<code>bolt-small</code>	Gurgaon	0.027	37.04	61.1	92.8	262.82
<code>t5-mini</code>	Patna	0.50	1.99	58.1	91.6	163.17
<code>t5-small</code>	Patna	0.78	1.28	57.4	92.3	157.03
<code>bolt-small</code>	Patna	0.025	40.73	55.0	92.9	117.11

## 5.4 Visualizations

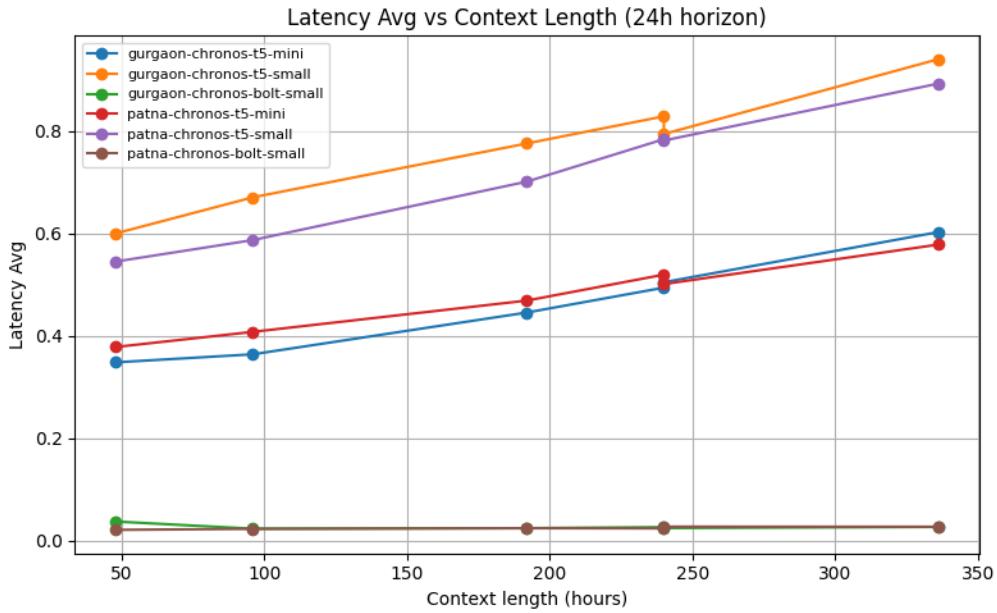


Figure 5: Average latency vs. context length.

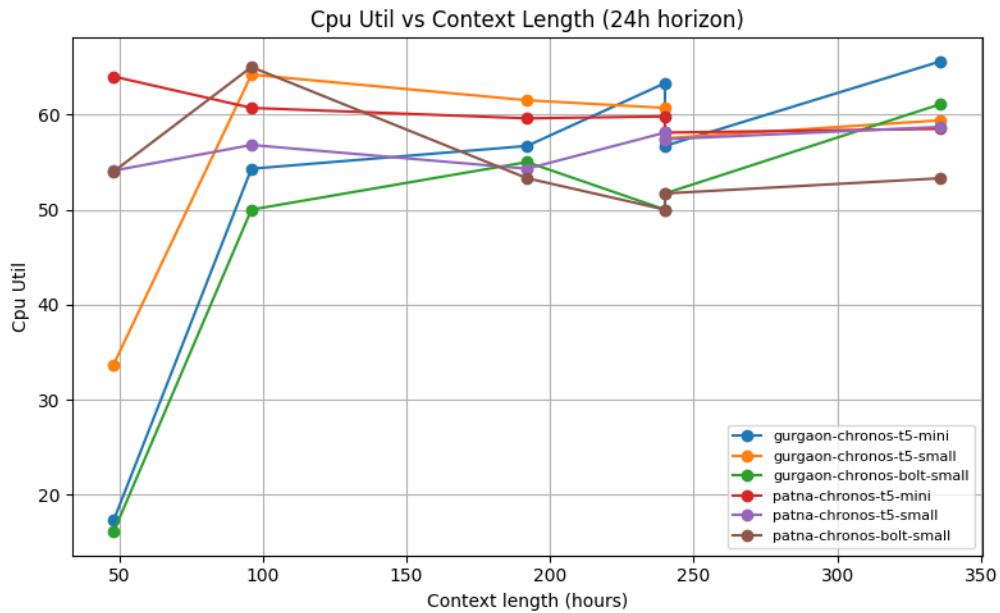


Figure 6: CPU utilization vs. context length.