IQR Method Refinement

1. Skew-Adjusted IQR Thresholding

The standard IQR was calculated for each sensor variable to determine outliers. To enhance detection accuracy, skew-adjusted bounds were applied, taking into account each feature's skewness, which dynamically adjusted the thresholds.

Skewness was calculated using

which measures the asymmetry in the data distribution for each sensor reading:

- Positive skew: Indicates a longer or fatter tail on the right side of the distribution.
- **Negative skew**: Indicates a longer or fatter tail on the left side.
- **Zero or near-zero skew**: Suggests a symmetric distribution.

Parameter Values:

- Multiplier (k): 1.5, adjusted based on skewness for each feature.
- **Skew Adjustment:** For each feature, the multiplier was modified based on skewness as follows:

```
Lower Multiplier= 1.5 * (1 - skew)

Upper Multiplier= 1.5 * (1 + skew)
```

• This adjustment creates skew-sensitive bounds, refining outlier detection.

Outlier Bounds Calculation:

```
Lower Bound = Q1 - (Lower Multiplier * IQR)

Upper Bound = Q3 + (Upper Multiplier * IQR)
```

2. Results and Observations

Effect of Skew Adjustment:

Features with higher skewness, such as [specific feature with skew], had their
multipliers adjusted, resulting in tighter or wider bounds based on the direction
of skew. This led to a reduction in false positives and more accurately identified
true anomalies.

Outlier Detection Patterns:

- The skew-adjusted IQR method effectively detected genuine anomalies, particularly in high-variance accelerometer features like Acc Z.
- Features with lower skew, such as gyro_x, had minimal adjustment, retaining traditional IQR bounds with effective detection results.

Conclusion

The skew-adjusted IQR method, with a base multiplier of 1.5, effectively balanced sensitivity and specificity in outlier detection. This method reduced false positives, particularly for features with inherent skewness, making it suitable for high-variance, sensor-based anomaly detection tasks.