

# Statistical Outlier Detection Notes

## Outlier Detection Using Mean and Standard Deviation (Z-Score Based Outlier Detection)

To detect outliers in a dataset  $\Delta$ , we use the mean and standard deviation:

- $\mu(\Delta)$ : Mean of the data
- $\sigma(\Delta)$ : Standard deviation of the data

### Normal Range

The normal range is defined as:

$$\mu(\Delta) \pm 2\sigma(\Delta)$$

This means most data points (about 95% if normally distributed) are expected to lie within this range.

### Outlier Condition

A value is considered an outlier if:

$$\Delta < \mu(\Delta) - 2\sigma(\Delta) \text{ or } \Delta > \mu(\Delta) + 2\sigma(\Delta)$$

- $\Delta$  - Orderbook Delta Depth of 5% from Coinbase
- $\mu(\Delta)$  - Mean of  $\Delta$
- $\sigma(\Delta)$  - Standard deviation of the dataset

## Defining Z-score

### Python Equivalent

This logic can be implemented in Python as follows:

```
mu, std = delta.mean(), delta.std()
lower, upper = mu - 2*std, mu + 2*std
outliers = (delta < lower) | (delta > upper)
```

### Notes

- This method assumes data is roughly normally distributed.
- Using  $2\sigma$  captures approximately 95% of data points under a normal distribution.
- You can adjust the multiplier (e.g.,  $3\sigma$ ) for stricter or looser thresholds.