

A-1

ERROR ANALYSIS

**MOVING AVERAGE AND
SMOOTHING TECHNIQUES**



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Process or Product Monitoring and Control

4. Introduction to Time Series Analysis

6.4.2. What are Moving Average or Smoothing Techniques?

Smoothing data removes random variation and shows trends and cyclic components

Inherent in the collection of data taken over time is some form of random variation. There exist methods for reducing or canceling the effect due to random variation. An often used technique in industry is "smoothing". This technique, when properly applied, reveals more clearly the underlying trend, seasonal and cyclic components.

There are two distinct groups of smoothing methods

- Averaging Methods
- Exponential Smoothing Methods

Taking averages is the simplest way to smooth data

We will first investigate some averaging methods, such as the "simple" average of all past data.

A manager of a warehouse wants to know how much a typical supplier delivers in 1000 dollar units. He/she takes a sample of 12 suppliers, at random, obtaining the following results:

| Supplier | Amount | Supplier | Amount |
|----------|--------|----------|--------|
| 1 | 9 | 7 | 11 |
| 2 | 8 | 8 | 7 |
| 3 | 9 | 9 | 13 |
| 4 | 12 | 10 | 9 |
| 5 | 9 | 11 | 11 |
| 6 | 12 | 12 | 10 |

The computed mean or average of the data = 10. The manager decides to use this as the estimate for *expenditure of a typical supplier*.

Is this a good or bad estimate?

Mean squared error is a

We shall compute the "mean squared error":

- The "error" = true amount spent minus the estimated

*way to judge
how good a
model is*

amount.

- The "error squared" is the error above, squared.
- The "SSE" is the sum of the squared errors.
- The "MSE" is the Mean of the squared errors.

*MSE results
for example*

The results are:

Error and Squared Errors

The estimate = 10

| Supplier | \$ | Error | Error Squared |
|----------|----|-------|---------------|
| 1 | 9 | -1 | 1 |
| 2 | 8 | -2 | 4 |
| 3 | 9 | -1 | 1 |
| 4 | 12 | 2 | 4 |
| 5 | 9 | -1 | 1 |
| 6 | 12 | 2 | 4 |
| 7 | 11 | 1 | 1 |
| 8 | 7 | -3 | 9 |
| 9 | 13 | 3 | 9 |
| 10 | 9 | -1 | 1 |
| 11 | 11 | 1 | 1 |
| 12 | 10 | 0 | 0 |

The SSE = 36 and the MSE = $36/12 = 3$.

*Table of
MSE results
for example
using
different
estimates*

So how good was the estimator for the amount spent for each supplier? Let us compare the estimate (10) with the following estimates: 7, 9, and 12. That is, we estimate that each supplier will spend \$7, or \$9 or \$12.

Performing the same calculations we arrive at:

| Estimator | 7 | 9 | 10 | 12 |
|-----------|-----|----|----|----|
| SSE | 144 | 48 | 36 | 84 |
| MSE | 12 | 4 | 3 | 7 |

The estimator with the smallest MSE is the best. It can be shown mathematically that the estimator that minimizes the MSE for a set of random data is the mean.

Table
showing
squared
error
or for the
mean for
sample data

Next we will examine the mean to see how well it predicts net income over time.

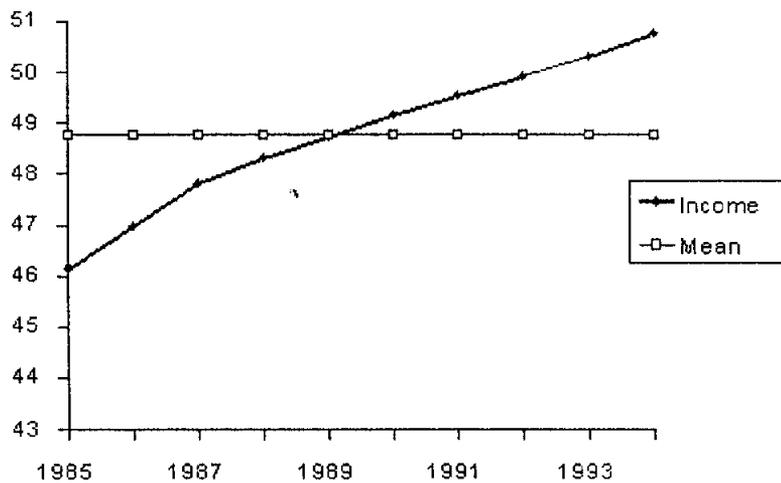
The next table gives the income before taxes of a PC manufacturer between 1985 and 1994.

| Year | \$ (millions) | Mean | Error | Squared Error |
|------|------------------|--------|--------|------------------|
| 1985 | 46.163 | 48.776 | -2.613 | 6.828 |
| 1986 | 46.998 | 48.776 | -1.778 | 3.161 |
| 1987 | 47.816 | 48.776 | -0.960 | 0.922 |
| 1988 | 48.311 | 48.776 | -0.465 | 0.216 |
| 1989 | 48.758 | 48.776 | -0.018 | 0.000 |
| 1990 | 49.164 | 48.776 | 0.388 | 0.151 |
| 1991 | 49.548 | 48.776 | 0.772 | 0.596 |
| 1992 | 48.915 | 48.776 | 1.139 | 1.297 |
| 1993 | 50.315 | 48.776 | 1.539 | 2.369 |
| 1994 | 50.768 | 48.776 | 1.992 | 3.968 |

The MSE = 1.9508

The mean is
not a good
estimator
when there
are trends

The question arises: *can we use the mean to forecast income if we suspect a trend?* A look at the graph below shows clearly that we should not do this.



Average
weights all
past

In summary, we state that

1. The "simple" average or mean of all past observations is

*observations
equally*

only a useful estimate for forecasting when there are no trends. If there are trends, use different estimates that take the trend into account.

- The average "weighs" all past observations equally. For example, the average of the values 3, 4, 5 is 4. We know, of course, that an average is computed by adding all the values and dividing the sum by the number of values. Another way of computing the average is by adding each value divided by the number of values, or

$$3/3 + 4/3 + 5/3 = 1 + 1.3333 + 1.6667 = 4.$$

The multiplier $1/3$ is called the *weight*. In general:

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i = \left(\frac{1}{n}\right) x_1 + \left(\frac{1}{n}\right) x_2 + \dots + \left(\frac{1}{n}\right) x_n$$

The $\left(\frac{1}{n}\right)$ are the weights and of course they sum to 1.

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