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# Chapter 1

## Bland - Altman Methodology

### 1.1 Bland Altman Plots In Literature

? contains a study the use of Bland Altman plots of 44 articles in several named journals over a two year period. 42 articles used Bland Altman's limits of agreement, wit the other two used correlation and regression analyses. ? remarks that 3 papers, from 42 mention predefined maximum width for limits of agreement which would not impair medical care.

The conclusion of ? is that there are several inadequacies and inconsistencies in the reporting of results ,and that more standardization in the use of Bland Altman plots is required. The authors recommend the prior determination of limits of agreement before the study is carried out. This contention is endorsed by ?, which makes a similar recommendation for the sample size, noting that *sample sizes required either was not mentioned or no rationale for its choice was given.*

### 1.2 The Bland Altman Plot

In 1986 Bland and Altman published a paper in the Lancet proposing the difference plot for use for method comparison purposes. It has proved highly popular ever since. This is a simple, and widely used , plot of the differences of each data pair, and the

corresponding average value. An important requirement is that the two measurement methods use the same scale of measurement.

Variations of the Bland Altman plot is the use of ratios, in the place of differences.

$$D_i = X_i - Y_i \tag{1.1}$$

Altman and Bland suggest plotting the within subject differences  $D = X_1 - X_2$  on the ordinate versus the average of  $x_1$  and  $x_2$  on the abscissa.

### 1.3 Criticism of Bland Altman Plot

Unfortunately the Bland-Altman plot has a fatal flaw: it indicates incorrectly that there are systematic differences or bias in the relationship between two measures, when one has been calibrated against the other. (Hopkins)

### 1.4 Bland Altman Plots

The issue of whether two measurement methods are comparable to the extent that they can be used interchangeably with sufficient accuracy is encountered frequently in scientific research. Historically comparison of two methods of measurement was carried out by use of matched pairs correlation coefficients or simple linear regression. Bland and Altman recognized the inadequacies of these analyses and articulated quite thoroughly the basis on which of which they are unsuitable for comparing two methods of measurement (?).

As an alternative they proposed a simple statistical methodology specifically appropriate for method comparison studies. They acknowledge that there are other valid methodologies, but argue that a simple approach is preferable to complex approaches, *“especially when the results must be explained to non-statisticians”* (?).

The first step recommended, which the authors argue should be mandatory, is construction of a simple scatter plot of the data. The line of equality ( $X = Y$ ) should also be shown, as it is necessary to give the correct interpretation of how both methods

compare. A scatter plot of the Grubbs data is shown in figure 2.1. A visual inspection thereof confirms the previous conclusion that there is an inter method bias present, i.e. Fotobalk device has a tendency to record a lower velocity.

In light of shortcomings associated with scatterplots, ? recommend a further analysis of the data. Firstly differences of measurements of two methods on the same subject should be calculated, and then the average of those measurements (Table 1.1). The averages of the two measurements is considered by Bland and Altman to the best estimate for the unknown true value. Importantly both methods must measure with the same units. These results are then plotted, with differences on the ordinate and averages on the abscissa (figure 1.2).

The dashed line in figure 1.2 alludes to the inter method bias between the two methods, as mentioned previously. Bland and Altman recommend the estimation of inter method bias by calculating the average of the differences. In the case of Grubbs data the inter method bias is  $-0.61$  metres per second.

Round	Fotobalk [F]	Counter [C]	Differences [F-C]	Averages [(F+C)/2]
1	793.80	794.60	-0.80	794.20
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7	791.70	792.40	-0.70	792.00
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12	793.50	793.80	-0.30	793.60

Table 1.1: Fotobalk and Counter Methods: Differences and Averages

From a visual inspection of Bland-Altman plot, it is also possible to compare the precision of each method in addition to the inter-method bias. Evidently the data points in the figure 1.2 tend to cluster near the bias line, particularly at the lower end of the range of measurements. The variances of the differences seem to increase along the range. In case of small data sets, any decision on the level of precision is subjective.

### 1.4.1 Inspecting the Data

Bland-Altman plots are a powerful graphical methodology for making a visual assessment of the data. ? express the motivation for this plot thusly:

”From this type of plot it is much easier to assess the magnitude of disagreement (both error and bias), spot outliers, and see whether there is any trend, for example an increase in (difference) for high values. This way of plotting the data is a very powerful way of displaying the results of a method comparison study.”

Figures XX YY and ZZ are three Bland-Altman plots derived from simulated data, each for the purpose of demonstrating how the plot would inform an analyst of trends that would adversely affect use of the recommended methodology. Figure XX demonstrates how the Bland Altman plot would indicate increasing variance of differences over the measurement range.

Figure ZZ is an example of cases where the inter-method bias changes over the measurement range. This is known as proportional bias (?).

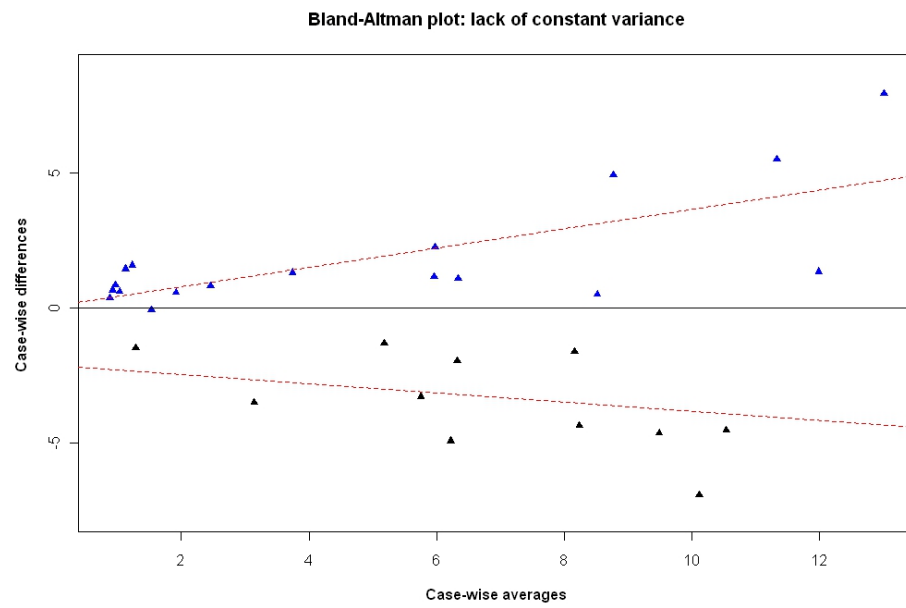


Figure 1.1: Bland-Altman Plot demonstrating the increase of variance over the range

Figure ZZ is an example of cases where the inter-method bias changes over the measurement range. This is known as proportional bias (Ludbrook, 1997). Both of these cases violate the assumptions necessary for further analysis using limits of agreement, which shall be discussed later. The plot also can be used to identify outliers. An outlier is an observation that is numerically distant from the rest of the data. Classification thereof is a subjective decision in any analysis, but must be informed by the logic of the formulation. Figure YY is a Bland Altman plot with two conspicuous observations, at the extreme left and right of the plot respectively.

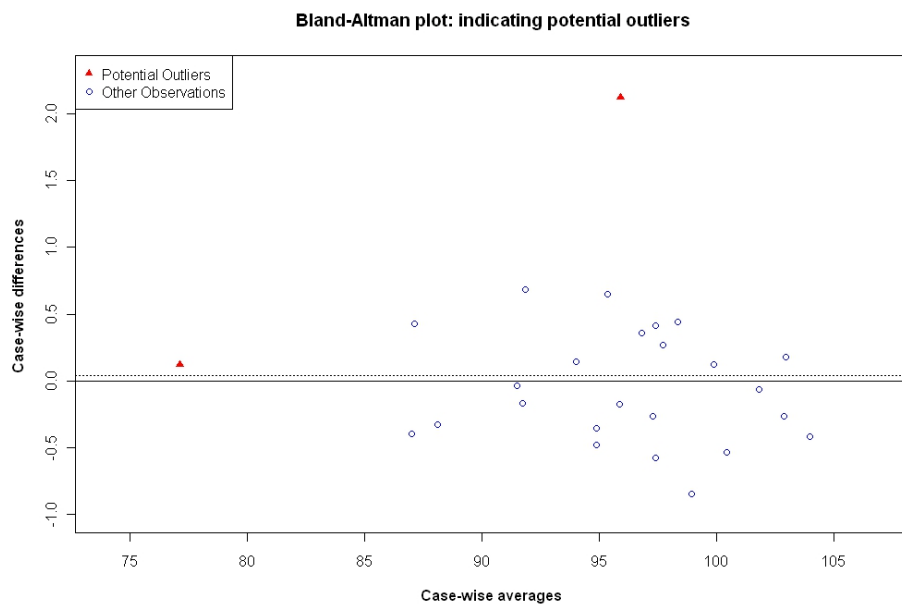


Figure 1.2: Bland-Altman Plot indicating the presence of Outliers

In the Bland-Altman plot, the horizontal displacement of any observation is supported by two independent measurements. Hence any observation, such as the one on the extreme right of figure YY, should not be considered an outlier on the basis of a noticeable horizontal displacement from the main cluster. The one on the extreme left should be considered an outlier, as it has a noticeable vertical displacement from the rest of the observations.



? do not recommend excluding outliers from analyses. However recalculation of the inter-method bias estimate , and further calculations based upon that estimate, are useful for assessing the influence of outliers.(?) states that "*We usually find that this method of analysis is not too sensitive to one or two large outlying differences.*"

## 1.5 Limits of Agreement

? introduces an elaboration of the plot, adding to the plot ‘limits of agreement’ to the plot. These limits are based upon the standard deviation of the differences. The discussion shall be reverted to these limits of agreement in due course.

### 1.5.1 Variations of the Bland Altman Plot

? remarks that it is possible to ignore the issue altogether, but the limits of agreement would wider apart than necessary when just lower magnitude measurements are considered. Conversely the limits would be too narrow should only higher magnitude measurements be used. To address the issue, they propose the logarithmic transformation of the data. The plot is then formulated as the difference of paired log values against their mean. ? acknowledge that this is not easy to interpret, and that it is not suitable in all cases.

? offers two variations of the Bland -Altman plot that are intended to overcome potential problems that the conventional plot would inappropriate for.

The first variation is a plot of casewise differences as percentage of averages, and is appropriate when there is an increase in variability of the differences as the magnitude increases.

## 1.6 Bland Altman Plots

The issue of whether two measurement methods are comparable to the extent that they can be used interchangeably with sufficient accuracy is encountered frequently in scientific research. Historically comparison of two methods of measurement was carried out by use of matched pairs correlation coefficients or simple linear regression. Bland and Altman recognized the inadequacies of these analyses and articulated quite thoroughly the basis on which of which they are unsuitable for comparing two methods of measurement (?).

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The first step recommended which the authors argue should be mandatory is construction of a simple scatter plot of the data. The line of equality ( $X = Y$ ) should also be shown, as it is necessary to give the correct interpretation of how both methods compare. A scatter plot of the Grubbs data is shown in figure 2.1. A visual inspection thereof confirms the previous conclusion that there is an inter method bias present, i.e. Fotobalk device has a tendency to record a lower velocity.

In light of shortcomings associated with scatterplots, ? recommend a further analysis of the data. Firstly differences of measurements of two methods on the same subject should be calculated, and then the average of those measurements (Table 1.1). The averages of the two measurements is considered by Bland and Altman to the best estimate for the unknown true value. Importantly both methods must measure with the same units. These results are then plotted, with differences on the ordinate and averages on the abscissa (figure 1.2). ?express the motivation for this plot thusly:

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From a visual inspection of Bland-Altman plot, it is also possible to compare the precision of each method in addition to the inter-method bias. Evidently the data points in the figure HH tend to cluster near the bias line, particularly at the lower end of the range of measurements. The variances of the differences seem to increase along the range. In case of small data sets, Any decision on the level of precision is subjective.

Figures XX and ZZ are Bland-Altman plots of data simulated for expository purposes. Figure XX demonstrates how the Bland Altman plot would indicate increasing variance of differences over the measurement range. Figure ZZ is an example of cases where the inter-method bias changes over the measurement range. This is known as proportional bias (?).

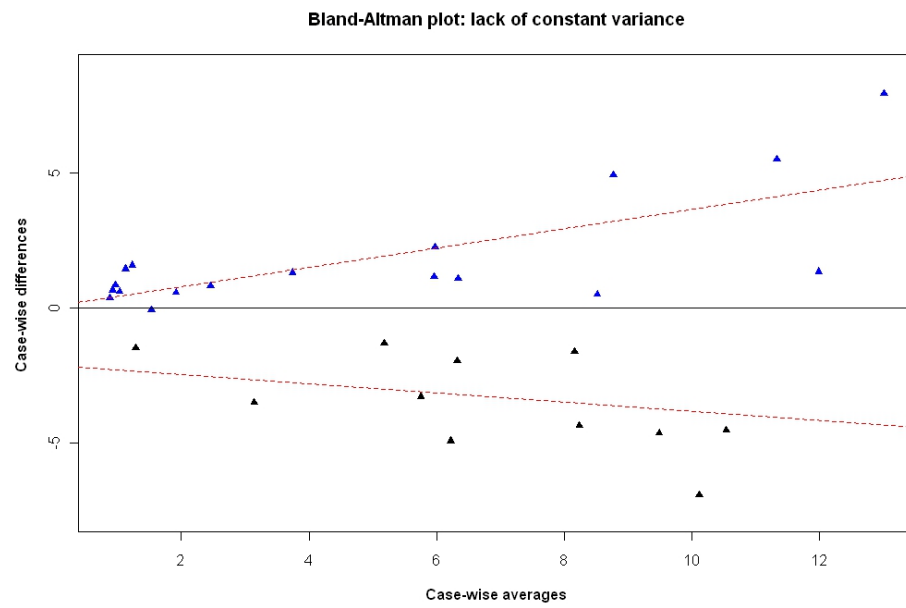


Figure 1.3: Bland-Altman Plot demonstrating the increase of variance over the range

The fourth pair of measurements from table 1.1 show both methods recording the same value, hence the difference is zero. In assessing the impact of the corresponding data point, two conflicting conclusions can be drawn.

One conclusion is that it is an outlier, and the precision of the differences is consistent along the range of measurements. The other conclusion is that it is not an outlier, and the precision increases proportionately along the range of measurements.

The Bland Altman plot is a simple tool for inspection of the data, but in itself it offers no formal testing procedure in this regard. To this end, the approach proposed by ? is a formal test on the Pearson correlation coefficient of casewise differences and means ( $\rho_{AD}$ ). According to the authors, this test is equivalent to a well established tests for equality of variances, known as the ‘Pitman Morgan Test’ (??).

For the Grubbs data, the correlation coefficient estimate ( $r_{AD}$ ) is 0.2625, with a 95% confidence interval of (-0.366, 0.726) estimated by Fishers ‘r to z’ transformation (?). The null hypothesis ( $\rho_{AD} = 0$ ) would fail to be rejected. Consequently the null hypothesis of equal variances of each method would also fail to be rejected.

There has been no further mention of this particular test in the subsequent article published by Bland and Altman, although ? refers to Spearmans’ rank correlation coefficient.

The second variation is a plot of casewise ratios as percentage of averages.

## 1.7 Bland Altman plot and the Treatment of Outliers

We wish to determine how outliers should be treated in a Bland Altman Plot. In their 1983 paper Bland and Altman merely state that the plot can be used to ‘spot outliers’. However they pay more attention to the issue of outliers in their 1986 paper, wherein they present a data set with an extreme outlier.

In light of shortcomings associated with scatterplots, ? recommend a further analysis of the data. Firstly differences of measurements of two methods on the same

subject should be calculated, and then the average of those measurements (Table 1.1). The averages of the two measurements is considered by Bland and Altman to the best estimate for the unknown true value. Importantly both methods must measure with the same units. These results are then plotted, with differences on the ordinate and averages on the abscissa (figure 1.2). Express the motivation for this plot thusly:

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There has been no further mention of this particular test in the subsequent article published by Bland and Altman, although ? refers to Spearmans’ rank correlation coefficient.

### 1.7.1 Criticism of Bland Altman Plot

Hopkins[8] argues that the plot indicates incorrectly that there are systematic differences or bias in the relationship between two measures, when one has been calibrated against the other.

An Evaluation of the correlation between the difference and means complement the analysis.

Bland and Altman caution that the calculations are based on the assumption that the data is normally distributed. This can be verified by using a histogram. If Data is not

normally distributed, it can be transformed.

### 1.7.2 Treatment of Outliers

Bland and Altman attend to the issue of outliers in their 1986 paper, wherein they present a data set with an extreme outlier

## 1.8 Bland Altman plot and the Treatment of Outliers

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In Bland and Altman's 1999 paper, we get the clearest indication of what Bland and Altman suggest on how to react to the presence of outliers. Their recommendation is to recalculate the limits without them, in order to test the difference with the calculation where outliers are retained. *The span has reduced from 77 to 59 mmHg, a noticeable but not particularly large reduction.* However, they do not recommend removing outliers. Furthermore, they say: *We usually find that this method of analysis is not too sensitive to one or two large outlying differences.*

In their 1986 paper, Bland and Altman give an example of an outlier. They state that it could be omitted in practice, but make no further comments on the matter.

We ask if this would be so in all cases. Given that the limits of agreement may or may not be disregarded, depending on their perceived suitability, we examine whether it would be possible that the deletion of an outlier may lead to a calculation of limits of agreement that are usable in all cases?

Should an Outlying Observation be omitted from a data set? In general, this is not considered prudent. Also, it may be required that the outliers are worthy of particular

attention themselves.

## 1.9 Variations of the Bland Altman Plot

? offers two variations of the Bland -Altman plot that are intended to overcome potential problems that the conventional plot would inappropriate for.

The first variation is a plot of casewise differences as percentage of averages, and is appropriate when there is an increase in variability of the differences as the magnitude increases.

The second variation is a plot of casewise ratios as percentage of averages.

## **Classifying outliers and recalculating**

We opted to examine this matter in more detail. The following points have to be considered:

How to suitably identify an outlier (in a generalized sense)

Would a recalculation of the limits of agreement generally results in a compacted range between the upper and lower limits of agreement?

## 1.10 Formal Testing

The Bland-Altman plot is a simple tool for inspection of the data, but in itself it offers no formal testing procedure in this regard. To this end, the approach proposed by ? is a formal test on the Pearson correlation coefficient of casewise differences and means ( $\rho_{AD}$ ). According to the authors, this test is equivalent to a well established tests for equality of variances, known as the ‘Pitman Morgan Test’ (??).

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