

## Seminar 4 - Hat Value Math 567: Winter 2016

### 1 Little Background

### 2 Highly influential data point

A data point is said to be influential if when removed from the calculation change the regression line significantly. How influential a data-point is, is the combination of how much leverage it has and how extreme it is in the  $y$  direction. However, A data point can have an high leverage but not influential, and it goes the same way for an outlier(all outlier are not influential).

### 3 Definition

Hat-matrix

The algebraic expression.

$$\hat{y} = Hy$$

Where:

$\hat{Y}_j$  is the prediction from the full regression model for observation  $j$ ;

$\hat{Y}_{j(i)}$  is The prediction for observation  $j$  from a refitted regression model in which observation  $i$  has been omitted;

The **leverage** define how far apart is a given data point from the average(mean/median). Points with high leverage tend to pull the regression line toward themselves and have impact on the slop of the regression line hence **influential**.

### 4 Interpretation of Hat values

There are several rules when interpreting **cook's distance**. The widely used criterion is that a point is considered to be highly influential if  $D_i > 1$  [?]

Different rules have been defined such as:  $D_i > 8.5$  if  $p > 3$  [?] where  $p$  is the number of regression parameter. [?] declares a data-point to be influential when  $D_i > \frac{4}{n}$  where  $n$  is the number of observation.

## 5 Hat values using R

Packages use:

```
install.packages (QuantPshyc)
```

```
library(QuantPshyc)
```

call Cook's  $D_i$  from the library: `linearmodel.cook()`

manually computing cook's distance:

## 6 Discussion

What to do when a given data-point's  $D_i > 1$  ? has an