

# Robust Regression and Fama-French 1992 Revisited

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# Outline

## 1 Introduction

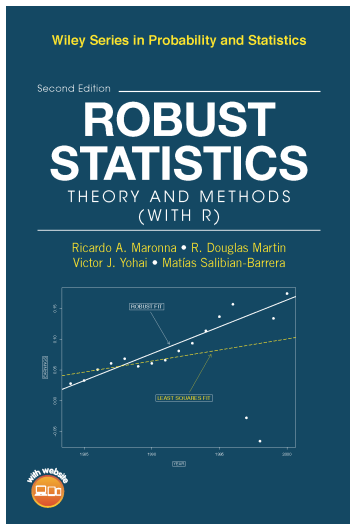
# Main Reference and R Package

**Robust Statistics: Theory and Methods (2019)**, 2nd ed.  
Maronna, Martin, Yohai and Salibian-Barrera (MMYS)

## Companion R Package: RobStatTM

To install, load, and view functions and data sets from R:

```
> install.packages("RobStatTM")  
> library(RobStatTM)  
> ls("package:RobStatTM")  
> data(package = "RobStatTM")
```



## Parametric Models Focus

Location, scale, and linear regression models

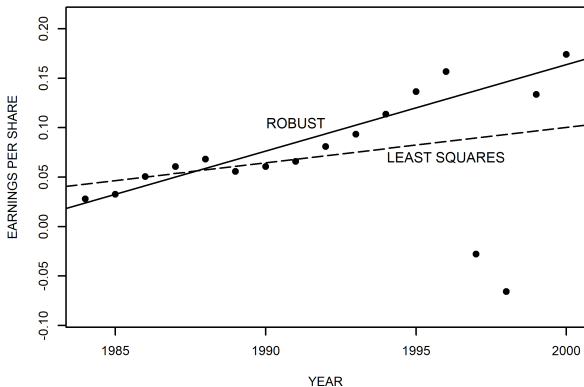
### 1 Data oriented definition of a robust estimator

- Parameter estimates are not much influenced by outliers
- Results in a good fit to the bulk of the data
- Reliable detection of multivariate outlier
- Confidence intervals and tests are not much influenced by outliers

### 2 Theoretical foundation

- Good estimator performance at normal and “nearly normal” distributions
- Universal use of two component nearly normal mixture distributions
- Performance measures: asymptotic variance and bias
- Supported by extensive finite sample mean-square error Monte Carlo

# Simple Example



- Robust regression provides a good fit to the bulk of the data
- Most robust regression residuals are smaller than for LS, with two robust residuals are larger than for LS

# Huber Location M-estimators (1964)

## Location and Scale Model

$$r_t = \mu + s \cdot \epsilon_t, \quad t = 1, 2, \dots, T, \quad \text{i.i.d. } \epsilon_t \stackrel{d}{\sim} f_0$$

## M-estimator:

$$\hat{\mu} = \underset{\mu}{\operatorname{argmin}} \sum_{t=1}^n \rho \left( \frac{r_t - \mu}{s} \right)$$

$$\sum_{t=1}^n \psi \left( \frac{r_t - \hat{\mu}}{s} \right) = 0, \quad \psi = \rho'$$

**MLE:**  $\rho = -\log f_0$