Definitions of ψ -Functions Available in Robustbase

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Preamble

Unless otherwise stated, the following definitions of functions are given by Maronna et al. (2006, p. 31), however our definitions differ sometimes slightly from theirs, as we prefer a different way of *standardizing* the functions. To avoid confusion, we first define ψ - and ρ -functions.

Definition 1 A ψ -function is a piecewise continuous function $\psi: \mathbb{R} \to \mathbb{R}$ such that

- 1. ψ is odd, i.e., $\psi(-x) = -\psi(x) \forall x$,
- 2. $\psi(x) \ge 0$ for $x \ge 0$, and $\psi(x) > 0$ for $0 < x < x_r := \sup\{\tilde{x} : \psi(\tilde{x}) > 0\}$ $(x_r > 0, possibly x_r = \infty)$.
- 3^* Its slope is 1 at 0, i.e., $\psi'(0) = 1$.

Note that '3*' is not strictly required mathematically, but we use it for standardization in those cases where ψ is continuous at 0. Then, it also follows (from 1.) that $\psi(0) = 0$, and we require $\psi(0) = 0$ also for the case where ψ is discontinuous in 0, as it is, e.g., for the M-estimator defining the median.

Definition 2 A ρ -function can be represented by the following integral of a ψ -function,

$$\rho(x) = \int_0^x \psi(x)dx \,, \tag{1}$$

which entails that $\rho(0) = 0$ and ρ is an even function.

A ψ -function is called redescending if $\psi(x) = 0$ for all $x \geq x_r$ for $x_r < \infty$. Corresponding to a redescending ψ -function, we define the function $\tilde{\rho}$, a version of ρ standardized such as to attain maximum value one. Formally,

$$\tilde{\rho}(x) = \rho(x)/\rho(\infty). \tag{2}$$

Note that $\rho(\infty) = \rho(x_r) \equiv \rho(x) \ \forall |x| >= x_r$. $\tilde{\rho}$ is a ρ -function as defined in Maronna et al. (2006) and has been called χ function in other contexts.

1 Monotone ψ -Functions

Montone ψ -functions lead to convex ρ -functions such that the corresponding M-estimators are defined uniquely.

Historically, the "Huber function" has been the first ψ -function, proposed by Peter Huber in Huber (1964).

1.1 Huber

The family of Huber functions is defined as,

$$\rho_k(x) = \begin{cases} \frac{1}{2}x^2 & \text{if } |x| \le k \\ k(|x| - \frac{k}{2}) & \text{if } |x| > k \end{cases},$$

$$\psi_k(x) = \begin{cases} x & \text{if } |x| \le k \\ k & \text{sign}(x) & \text{if } |x| > k \end{cases}.$$

The constant k for 95% efficiency of the regression estimator is 1.345.

> plot(huberPsi, x., ylim=c(-1.4, 5), leg.loc="topright", main=FALSE)

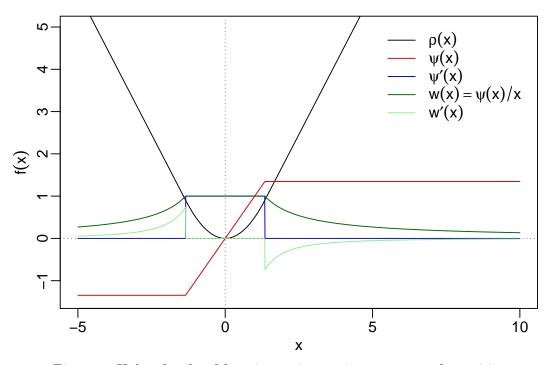


Figure 1: Huber family of functions using tuning parameter k = 1.345.

2 Redescenders

For the MM-estimators and their generalizations available via lmrob(), the ψ -functions are all redescending, i.e., with finite "rejection point" $x_r = \sup\{t; \psi(t) > 0\} < \infty$. From lmrob, the psi functions are available via lmrob.control,

> formals(lmrob.control) \$ psi

c("bisquare", "lqq", "welsh", "optimal", "hampel", "ggw")

and their ψ , ρ , ψ' , and weight function $w(x) := \psi(x)/x$, are all computed efficiently via C code, and are defined and visualized in the following subsections.

2.1 Bisquare

Tukey's bisquare (aka "biweight") family of functions is defined as,

$$\tilde{\rho}_k(x) = \begin{cases} 1 - (1 - (x/k)^2)^3 & \text{if } |x| \le k \\ 1 & \text{if } |x| > k \end{cases}$$

with derivative $\tilde{\rho}_k'(x) = 6\psi_k(x)/k^2$ where,

$$\psi_k(x) = x \left(1 - \left(\frac{x}{k} \right)^2 \right)^2 \cdot I_{\{|x| \le k\}} .$$

The constant k for 95% efficiency of the regression estimator is 4.685 and the constant for a breakdown point of 0.5 of the S-estimator is 1.548.

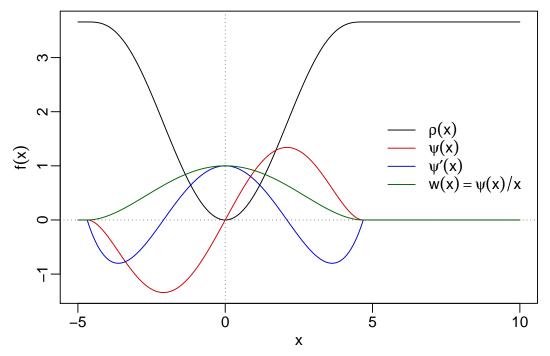


Figure 2: Bisquare family functions using tuning parameter k = 4.685.

2.2 Hampel

The Hampel family of functions (Hampel et al., 1986) is defined as,

$$\tilde{\rho}_{a,b,r}(x) = \begin{cases} \frac{1}{2}x^2/C & |x| \le a \\ \left(\frac{1}{2}a^2 + a(|x| - a)\right)/C & a < |x| \le b \\ \frac{a}{2}\left(2b - a + (|x| - b)\left(1 + \frac{r - |x|}{r - b}\right)\right)/C & b < |x| \le r \\ 1 & r < |x| \end{cases},$$

$$\psi_{a,b,r}(x) = \begin{cases} x & |x| \le a \\ a & \text{sign}(x) & a < |x| \le b \\ a & \text{sign}(x) \frac{r - |x|}{r - b} & b < |x| \le r \\ 0 & r < |x| \end{cases}$$

where $C := \rho(\infty) = \rho(r) = \frac{a}{2} (2b - a + (r - b)) = \frac{a}{2} (b - a + r)$.

As per our standardization, ψ has slope 1 in the center. The slope of the redescending part $(x \in [b, r])$ is -a/(r-b). If it is set to $-\frac{1}{2}$, as recommended sometimes, one has

$$r = 2a + b$$
.

Here however, we restrict ourselves to a = 1.5k, b = 3.5k, and r = 8k, hence a redescending slope of $-\frac{1}{3}$, and vary k to get the desired efficiency or breakdown point.

The constant k for 95% efficiency of the regression estimator is 0.9016 and the one for a breakdown point of 0.5 of the S-estimator is 0.212.

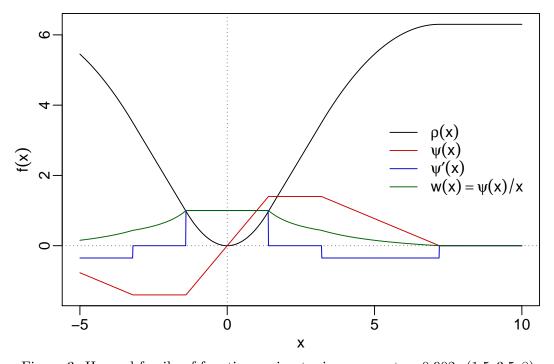


Figure 3: Hampel family of functions using tuning parameters $0.902 \cdot (1.5, 3.5, 8)$.

2.3 **GGW**

The Generalized Gauss-Weight function, or ggw for short, is a generalization of the Welsh ψ function (below). In Koller and Stahel (2011) it is defined as,

$$\psi_{a,b,c}(x) = \begin{cases} x & |x| \le c \\ \exp\left(-\frac{1}{2}\frac{(|x|-c)^b}{a}\right)x & |x| > c, \end{cases}.$$

The constants for 95% efficiency of the regression estimator are a=1.387, b=1.5 and c=1.063. The constants for a breakdown point of 0.5 of the S-estimator are a=0.204, b=1.5 and c=0.296.

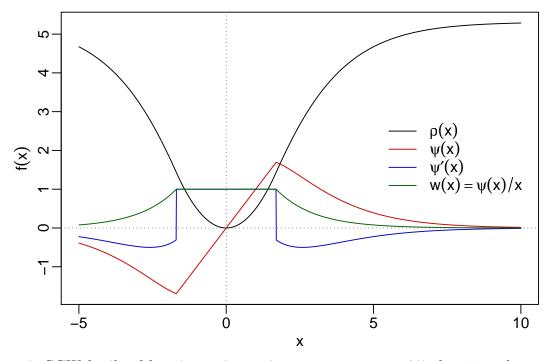


Figure 4: GGW family of functions using tuning parameters a = 1.387, b = 1.5 and c = 1.063.

2.4 LQQ

The "linear quadratic quadratic" ψ -function, or lqq for short, was proposed by Koller and Stahel (2011). It is defined as,

$$\psi_{b,c,s}(x) = \begin{cases} x & |x| \le c \\ \operatorname{sign}(x) \left(|x| - \frac{s}{2b} (|x| - c)^2 \right) & c < |x| \le b + c \\ \operatorname{sign}(x) \left(c + b - \frac{bs}{2} + \frac{s-1}{a} \left(\frac{1}{2} \tilde{x}^2 - a \tilde{x} \right) \right) & b + c < |x| \le a + b + c \\ 0 & \text{otherwise,} \end{cases}$$

where $\tilde{x} = |x| - b - c$ and a = (bs - 2b - 2c)/(1 - s). The parameter c determines the width of the central identity part. The sharpness of the bend is adjusted by b while the maximal rate of descent is controlled by s ($s = 1 - |\min_x \psi'(x)|$). The length a of the final descent to 0 is determined by b, c and s.

The constants for 95% efficiency of the regression estimator are $b=1.473,\ c=0.982$ and s=1.5. The constants for a breakdown point of 0.5 of the S-estimator are $b=0.402,\ c=0.268$ and s=1.5.

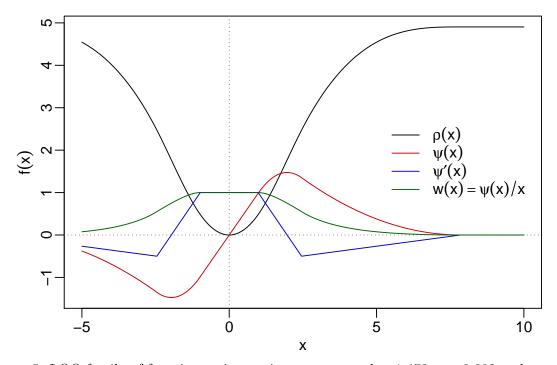


Figure 5: LQQ family of functions using tuning parameters b = 1.473, c = 0.982 and s = 1.5.

2.5 Optimal

The optimal ψ function as given by Maronna et al. (2006, Section 5.9.1),

$$\psi_c(x) = \operatorname{sign}(x) \left(-\frac{\varphi'(|x|) + c}{\varphi(|x|)} \right)_+,$$

where φ is the standard normal density, c is a constant and $t_+ := \max(t, 0)$ denotes the positive part of t.

The constant for 95% efficiency of the regression estimator is 1.060 and the constant for a breakdown point of 0.5 of the S-estimator is 0.405.

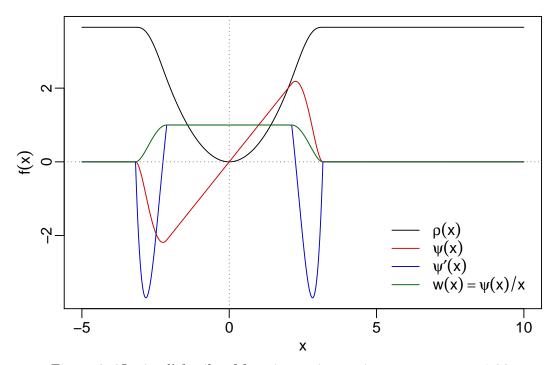


Figure 6: 'Optimal' family of functions using tuning parameter c = 1.06.

2.6 Welsh

The Welsh ψ function is defined as,

$$\tilde{\rho}_k(x) = 1 - \exp(-(x/k)^2/2)$$

$$\psi_k(x) = k^2 \tilde{\rho}'_k(x) = x \exp(-(x/k)^2/2)$$

$$\psi'_k(x) = (1 - (x/k)^2) \exp(-(x/k)^2/2)$$

The constant k for 95% efficiency of the regression estimator is 2.11 and the constant for a breakdown point of 0.5 of the S-estimator is 0.577.

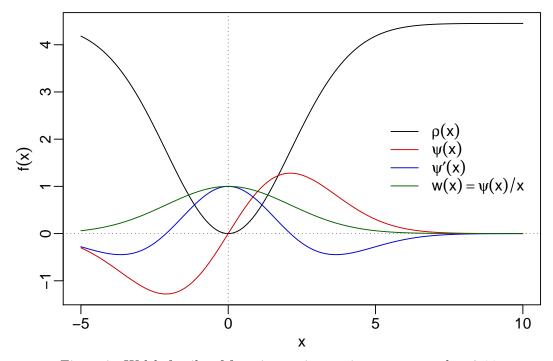


Figure 7: Welsh family of functions using tuning parameter k = 2.11.

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