



**Explicit models for bilateral fat-tailed distributions**  
and applications in finance with the package *FatTailsR*

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**R/Rmetrics Workshop - 27 June 2014**



A small company located in Paris:

- Consulting services and training
- Modeling and software design

Since 2009:

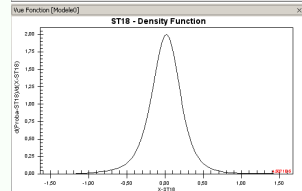
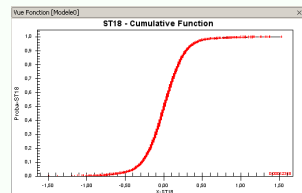
- Neural networks
- Design of experiments for nonlinear models

Since 2013:

- Bilateral fat-tailed distributions  
(from a work done with ST-Microelectronics)

June 2014:

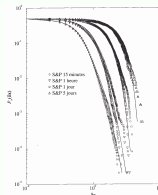
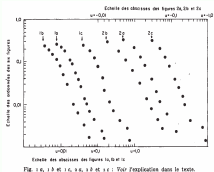
- Package **FatTailsR** to handle fat-tailed distributions in finance  
Models are symmetric (3 parameters) or asymmetric (4 parameters)  
Excellent results !!



- 1 Old models - New ideas
- 2 Symmetric models - Examples: Gold, Société Générale, Vivendi
- 3 Asymmetric models - Examples: S&P 500, Euro-Dollar, VIX
- 4 Package FatTailsR

# Fat-tailed bilateral distributions

- Almost all datasets related to stock returns over long periods
- Market risk (Solvency II, Bâle III), portfolio management, derivatives, ...
- Mentionned by **Mandelbrot (1962)**, reviewed by **Bouchaud and Potters (1997)**, who use a combination of unilateral models for left and right tails



$$\log \{ \text{Fr} [L(t, T) > u] \} \sim -\alpha \log u + \log C^+(T),$$

$$\log \{ \text{Fr} [L(t, T) < -u] \} \sim -\alpha \log u + \log C^-(T).$$

$$L_\mu(x) \simeq \frac{\mu A_\pm^\mu}{|x|^{1+\mu}} \text{ pour } x \rightarrow \pm\infty,$$

Figure : (a+b) Cotton price day-week-month - (c+d) Indice S&P 15min-day-week-month

- Centered and non-centered Student distributions
- Characteristic functions

⇒ We present a new, explicit, symmetric or asymmetric distribution

# Example: S&P 500

⇒ We present a new, explicit, symmetric or asymmetric distribution

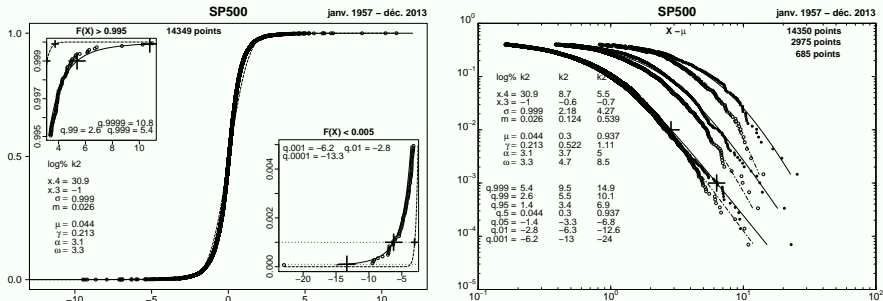


Figure : S&P 500 : (a) Cumulative function of the logreturns (b) Log-Log view day-week-month

Consider the logistic function which have thinner tails than the Laplace-Gauss function :

$$F(x) = \frac{1}{1 + e^{-x}}$$

It is remarkable that the combination of two asymmetric functions  $e^{-x}$  et  $\frac{1}{1 + \dots}$  gives a perfectly symmetric model. This comes from the fundamental property of the exponential  $e^{-x}e^x = 1$  which imposes  $F(-x) = 1 - F(x)$  and for the density function  $f(x) = F(x)F(-x)$

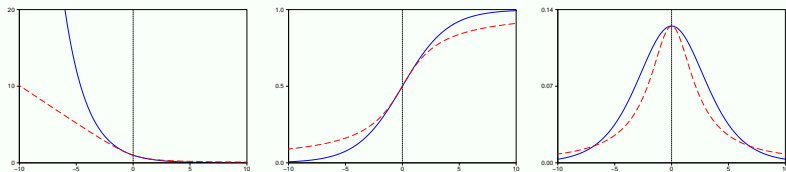


Figure : (a) Exp and hp (b) logis and logishp (c) dlogis and dlogishp

Consider the logistic function which have thinner tails than the Laplace-Gauss function :

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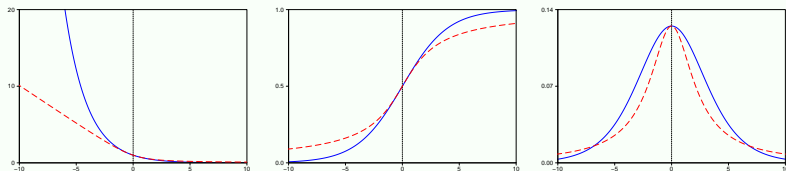


Figure : (a) Exp and hp (b) logis and logishp (c) dlogis and dlogishp

We use the other curves that verify the property  $y(-x)y(x) = 1$ :

- hyperbolas
- Power hyperbolas with parameter  $\kappa$  that allow the design of logistic type functions with tail convergence similar to  $|x|^{-\kappa}$

# Power hyperbolas

We call *power hyperbolas* the positive functions that verify:

## Generic equation of power hyperbolas

$$\left(y^{1/\kappa} + \frac{X - \mu}{\gamma\kappa}\right) y^{1/\kappa} = 1 \quad (1)$$

Solution to the equation are the curves:

$$y(X, \mu, \gamma, \kappa) = \left(-\frac{X - \mu}{2\gamma\kappa} + \sqrt{\left(\frac{X - \mu}{2\gamma\kappa}\right)^2 + 1}\right)^\kappa = e^{\kappa \log\left(-\frac{X - \mu}{2\gamma\kappa} + \sqrt{\left(\frac{X - \mu}{2\gamma\kappa}\right)^2 + 1}\right)}$$

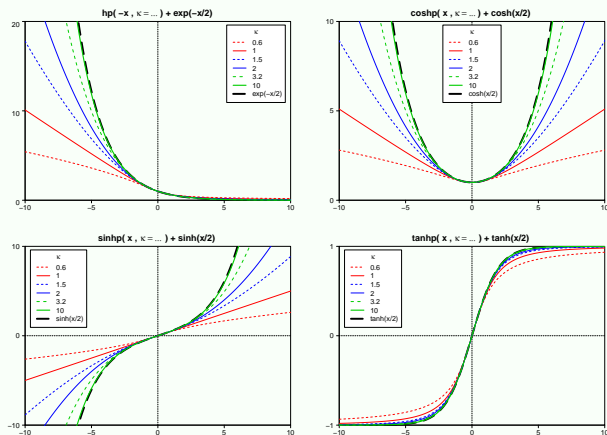
or:

## Power hyperbolas

$$y(X, \mu, \gamma, \kappa) = e^{-\kappa \operatorname{asinh}\left(\frac{X - \mu}{2\gamma\kappa}\right)} \xrightarrow{\kappa \rightarrow +\infty} e^{-\frac{X - \mu}{2\gamma}} \quad (2)$$

When  $\kappa = 1$ , the curve is the simple hyperbola





- Figure :**
- (a) Power hyperbola -  $\text{exp hp } \kappa(-x) = e^{-\kappa \operatorname{asinh}(\frac{x}{2\kappa})} \xrightarrow{\kappa \rightarrow +\infty} e^{-\frac{x}{2}}$
  - (b) Power hyperbolic cosine -  $\text{cosh p } \kappa(x) = \cosh \left[ \kappa \operatorname{asinh}(\frac{x}{2\kappa}) \right] \xrightarrow{\kappa \rightarrow +\infty} \cosh(\frac{x}{2})$
  - (c) Power hyperbolic sine -  $\text{sinh p } \kappa(x) = \sinh \left[ \kappa \operatorname{asinh}(\frac{x}{2\kappa}) \right] \xrightarrow{\kappa \rightarrow +\infty} \sinh(\frac{x}{2})$
  - (d) Power hyperbolic tangent -  $\text{tanh p } \kappa(x) = \tanh \left[ \kappa \operatorname{asinh}(\frac{x}{2\kappa}) \right] \xrightarrow{\kappa \rightarrow +\infty} \tanh(\frac{x}{2})$

# Power hyperbolas, cumulative functions, densities K1

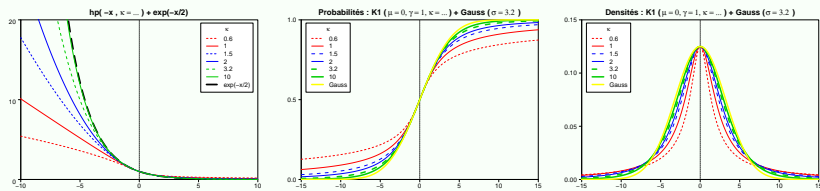


Figure : (a) Power hyperbolas - (b) Cumulative functions - (c) Densities

K1 model: cumulative functions and densities,  $T = \operatorname{asinh}\left(\frac{X-\mu}{2\gamma\kappa}\right)$ :

$$F(X) = \frac{1}{1 + e^{-\kappa \operatorname{asinh}\left(\frac{X-\mu}{2\gamma\kappa}\right)}} \quad f(X) = \frac{1}{4\gamma \cosh(T) (1 + \cosh(\kappa T))} \quad f(0) = \frac{1}{8\gamma} \quad (3)$$

$F$  and  $f$  verify Karamata theorem related to slowly varying functions:

$$\lim_{x \rightarrow -\infty} \frac{x f(x)}{F(x)} = -\kappa \quad \text{et} \quad \lim_{x \rightarrow +\infty} \frac{x f(x)}{1-F(x)} = \kappa$$

Quantiles and densities (calculated from the probability):

$$X = \mu + 2\gamma\kappa \sinh\left(\frac{\operatorname{logit}(p)}{\kappa}\right) \quad f(X) = \frac{1}{2\gamma} \operatorname{sech}\left(\frac{\operatorname{logit}(p)}{\kappa}\right) p(1-p)$$

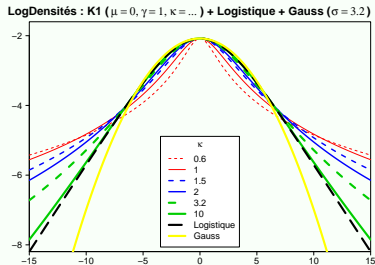
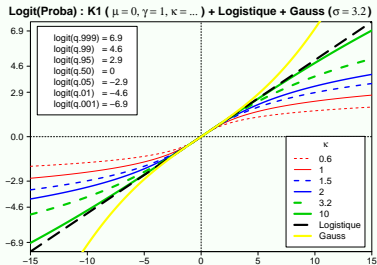


Figure : (a) QL-Plot = Quantiles + Logit of cumulative functions (b) Logdensities

$$\text{logit } F(X, \mu, \gamma, \kappa) = \kappa \operatorname{asinh} \left( \frac{X - \mu}{2\gamma\kappa} \right) \quad \log f(X) = -\log(8\gamma) - \log(\cosh(T)) - 2 \log \left( \cosh \left( \frac{\kappa T}{2} \right) \right)$$

Remarks on the QL-plot (a) with the logit of the cumulative function on y-axis:

- Logit-scale:  $\text{logit}(0.001, 0.01, 0.05, 0.95, 0.99, 0.999) = (-6.9, -4.6, -2.9, 0, 2.9, 4.6, 6.0)$
- The logit of the logistic function ( $\frac{x}{2}$ ) is the dashed bisecting line
- The Laplace-Gauss function (dashed lines+points) is concave and then convex
- The power hyperbola logistic functions are convex and then concave ( $\Rightarrow$  subexponential functions)
- The curvatures of the power hyperbola logistic functions depend on parameter  $\kappa$  and are well separated

# Comparison with Cauchy distribution ( $\kappa = 1$ )

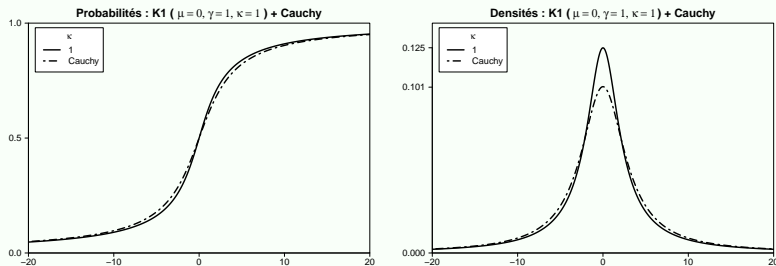


Figure : K1 model ( $\kappa = 1$ ) and Cauchy functions: (a) Cumulative functions - (b) Densities

$$F(x) = \frac{1}{1 - \frac{x}{2} + \sqrt{\left(\frac{x}{2}\right)^2 + 1}} \quad x = \frac{2p-1}{p(1-p)} \quad f(x) = \frac{1}{x^2 + 4 + 2\sqrt{x^2 + 4}} \quad f(x) = \frac{1}{\left(\frac{1}{p}\right)^2 + \left(\frac{1}{1-p}\right)^2}$$

$$G(x) = 0,5 + \frac{1}{\pi} \arctan\left(\frac{x}{\pi}\right) \quad x = \pi \tan(\pi(p - 0,5)) \quad g(x) = \frac{1}{\pi^2 + x^2} \quad g(x) = \frac{\cos^2(\pi(p - 0,5))}{\pi^2}$$

$$\lim_{X \rightarrow +\infty} F(X) = 1 - \frac{1}{x} + \frac{1}{x^2} - \frac{1}{x^4} + o\left(\frac{1}{x^5}\right) \quad \lim_{X \rightarrow +\infty} G(X) = 1 - \frac{1}{x} + \frac{\pi^2}{3x^3} - \frac{\pi^4}{5x^5} + o\left(\frac{1}{x^7}\right)$$

# Comparison with Laplace-Gauss distribution (any $\kappa$ , $\gamma = 0.313$ )

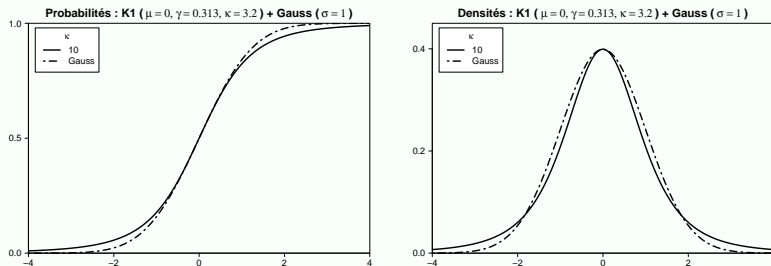


Figure : K1 model ( $\gamma = 0,313$ ,  $\kappa = 3.2$ ) and Laplace-Gauss distributions ( $\sigma = 1$ ): (a) Cumulative functions - (b) Densities

K1 and Laplace-Gauss distributions can be compared when they have same density peaks. The equality in  $X = \mu$  gives  $\tilde{h} = \frac{\zeta_\sigma}{\zeta_\gamma} = \frac{8\gamma}{\sqrt{2\pi}\sigma} = 1$ , or:

$$\gamma = \frac{\sqrt{2\pi}}{8} \sigma \approx 0,313 \sigma \quad \text{et} \quad \sigma = \frac{8}{\sqrt{2\pi}} \gamma \approx 3,192 \gamma$$

# Comparison with centered Student distribution

Centered Student distribution with  $\nu$  degrees of freedom:

$$f(x) = \frac{1}{\sqrt{\nu\pi}} \frac{\Gamma(\frac{\nu+1}{2})}{\Gamma(\frac{\nu}{2})} \left(1 + \frac{x^2}{\nu}\right)^{-\frac{\nu+1}{2}}$$

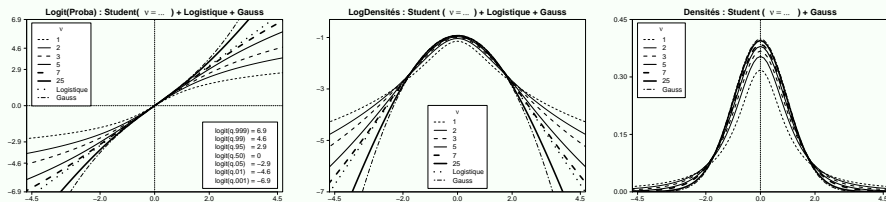


Figure : Student ( $\nu = 1, 2, 3, 5, 7, 25$ ), Logistic and Laplace-Gauss distributions:  
(a) Logit of cumulative functions - (b) Logdensities - (c) Densities

- $\nu$  can take integer values only:
  - $\nu < 1$  is impossible
  - $\nu = 1 \equiv$  Cauchy distribution.  $\nu = 2 \approx \kappa = 2$ .  $\nu = 8 \approx$  logistic distribution
  - $\nu = +\infty \equiv$  Laplace-Gauss
- The density peaks depend on  $\nu$
- In practice, one can use  $\nu = (2, 3, 4, 5, 6, 7, 8)$  which cover the subexponential domain

# Symmetric processes ( $\kappa = 2, \kappa = 3.2, \kappa = 5$ et $\kappa = 10$ ) x ( $\epsilon = 0$ )

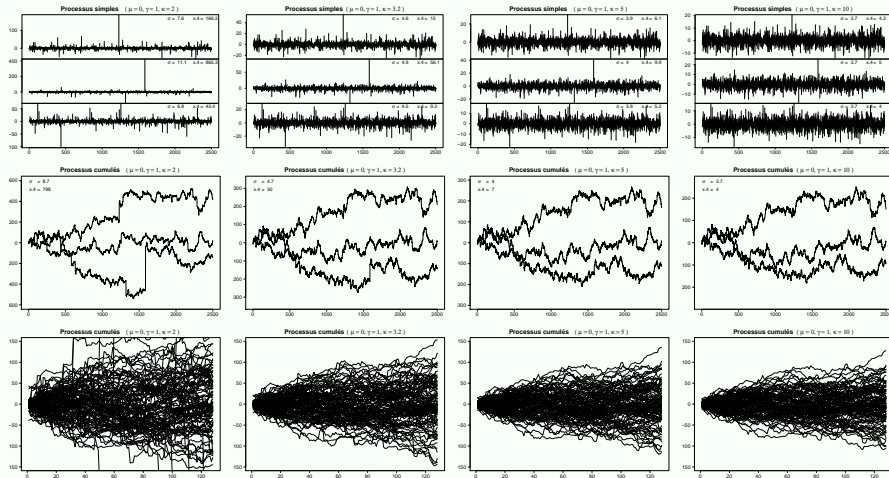
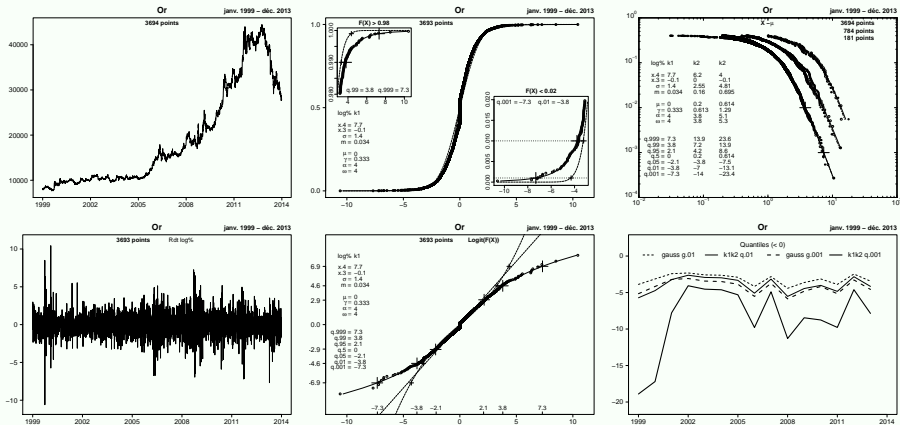


Figure : Processes and cumulated processes: (a)  $\kappa = 2$  (b)  $\kappa = 3.2$  (c)  $\kappa = 5$  (d)  $\kappa = 10$

# Gold lingot: $\kappa = 4$ , $\epsilon = 0$ , multi-scale symmetry



**Figure :** (a) Price of the gold lingot - (b) 100xlog-returns  
 (c) Cumulative function ( $\kappa = \alpha = \omega$ ) - (d) Logit of the cumulative function  
 (e) Cumulative function in Log-Log scale with periods day, week, month  
 (f) Risks at 1% at 1‰ over a yearly period (250 days) described by Laplace-Gauss and K1 estimates



# The annual risk profiles of Société Générale and Vivendi

The analysis of the returns can be conducted over fixed periods, for instance one year (about 250 days). The tail parameter  $\kappa$  describes the model curvature. q.001 is the risk at 1‰.

↓

Société-Générale	x.4	$\sigma$	m	$\mu$	$\gamma$	$\kappa$	q.001	q.01
1992	4.1	2.0	0.29	0.00	0.50	3.8	-11.5	-5.8
1993	3.3	1.4	0.09	0.13	0.38	12.0	-5.4	-3.4
1994	3.4	1.4	-0.12	-0.18	0.40	12.0	-5.9	-3.9
1995	4.9	1.7	0.03	-0.03	0.43	4.6	-8.5	-4.7
1996	3.9	1.3	-0.03	0.00	0.34	7.2	-5.5	-3.4
1997	4.3	1.9	0.15	0.21	0.51	5.4	-9.0	-5.1
1998	4.4	3.3	0.04	0.29	0.86	4.3	-17.1	-9.1
1999	6.0	2.5	0.20	0.23	0.60	3.3	-15.5	-7.2
2000	4.7	2.3	0.05	0.00	0.61	4.7	-11.8	-6.5
2001	4.6	2.3	-0.02	0.00	0.59	4.1	-12.6	-6.6
2002	4.6	3.4	-0.05	0.12	0.86	4.2	-18.1	-9.6
2003	4.3	2.1	0.09	0.00	0.57	5.3	-10.2	-5.9
2004	4.1	1.2	0.02	0.00	0.32	5.8	-5.6	-3.3
2005	3.9	1.0	0.13	0.11	0.28	6.5	-4.5	-2.6
2006	4.9	1.4	0.08	0.00	0.35	4.0	-7.7	-4.0
2007	4.7	1.8	-0.10	0.06	0.47	4.4	-9.4	-5.0
2008	<u>5.2</u>	<u>4.5</u>	<u>-0.37</u>	<u>-0.48</u>	<u>1.08</u>	<u>3.4</u>	<u>-27.8</u>	<u>-13.7</u>
2009	4.7	3.7	0.12	0.40	0.94	4.0	-20.0	-10.2
2010	<u>11.9</u>	3.1	-0.08	-0.10	0.67	2.8	-22.0	-9.4
2011	6.9	4.3	-0.33	-0.31	0.94	2.7	-32.6	-13.7
2012	3.7	3.3	0.20	0.20	0.91	12.0	-13.1	-8.4
2013	4.5	2.2	0.16	0.18	0.58	5.3	-10.3	-5.8
'92-'13	9.4	2.6	0.02	0.00	0.59	3.6	-14.2	-7.0

↑

↓

Vivendi	x.4	$\sigma$	m	$\mu$	$\gamma$	$\kappa$	q.001	q.01
1992	3.2	1.8	0.00	-0.08	0.52	12.0	-7.6	-4.9
1993	2.9	1.4	0.12	0.07	0.40	12.0	-5.8	-3.7
1994	3.6	1.8	-0.14	-0.20	0.49	12.0	-7.4	-4.8
1995	3.2	1.8	-0.02	-0.19	0.51	12.0	-7.6	-5.0
1996	4.5	1.4	0.11	0.00	0.35	4.5	-7.0	-3.8
1997	3.3	1.6	0.11	0.13	0.46	12.0	-6.6	-4.2
1998	4.5	1.9	0.22	0.16	0.51	4.9	-9.4	-5.2
1999	3.0	1.9	0.08	0.00	0.52	12.0	-7.6	-4.9
2000	4.2	2.7	-0.10	-0.06	0.72	5.0	-13.5	-7.6
2001	6.0	2.6	-0.05	-0.14	0.63	3.3	-16.6	-7.9
2002	<u>8.9</u>	<u>6.0</u>	<u>-0.54</u>	<u>-0.67</u>	<u>1.23</u>	<u>2.5</u>	<u>-49.0</u>	<u>-19.5</u>
2003	5.1	2.8	0.09	0.00	0.71	4.0	-15.3	-8.0
2004	4.2	1.5	0.08	0.05	0.40	5.0	-7.4	-4.2
2005	3.7	1.1	0.05	0.08	0.30	10.7	-4.3	-2.7
2006	3.2	1.1	0.04	0.11	0.32	12.0	-4.5	-2.9
2007	3.7	1.1	0.02	0.00	0.30	7.9	-4.7	-3.0
2008	6.8	2.5	-0.12	-0.04	0.59	3.1	-16.7	-7.6
2009	3.6	1.7	-0.04	0.00	0.48	12.0	-7.0	-4.5
2010	9.9	1.7	-0.01	0.08	0.36	2.7	-12.4	-5.1
2011	4.0	1.9	-0.07	0.03	0.52	6.3	-8.7	-5.2
2012	6.9	1.8	0.00	-0.02	0.46	3.8	-10.5	-5.3
2013	3.7	1.5	0.05	0.06	0.40	8.3	-6.2	-3.8
'92-'13	25.5	2.2	-0.01	0.00	0.45	2.9	-14.2	-6.1

↑



# The yearly return and risk profiles of Société Générale

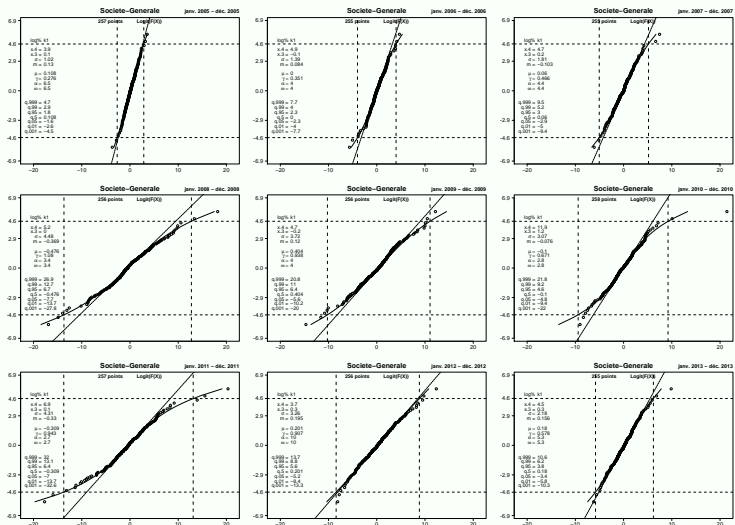
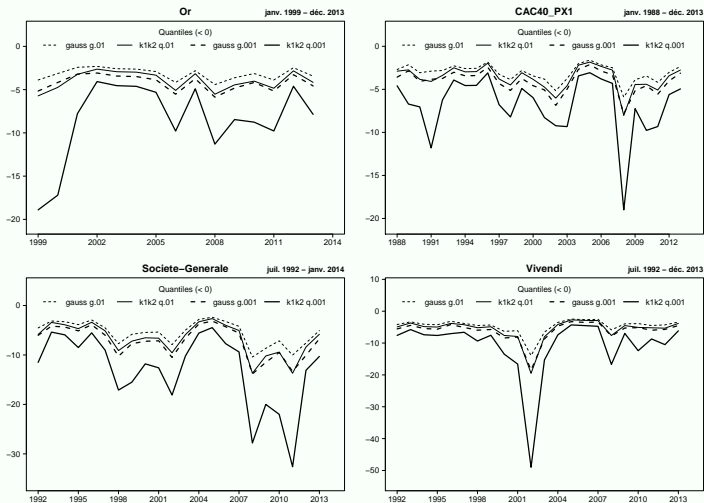


Figure : Logit of cumulative functions of Société Générale

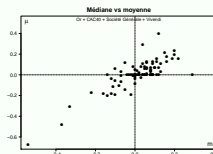
(a) 2005 - (b) 2006 - (c) 2007 - (d) 2008 - (e) 2009 - (f) 2010 - (g) 2011 - (h) 2012 - (i) 2013

# The yearly risk profiles

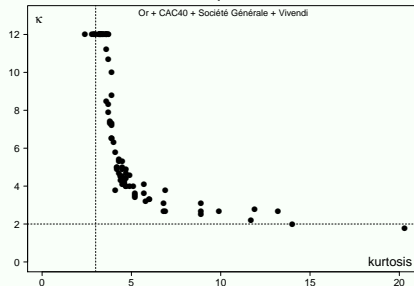


**Figure :** Quantiles: (a) Gold lingot - (b) CAC40 - (c) Société Générale - (d) Vivendi estimated with Laplace-Gauss (dashed lines) and K1 (full lines) models at levels 1% et 1‰ with datasets of about 250 points per year

# Comparison of K1 distribution parameters



Coefficient de queue vs kurtosis



Coefficient de queue vs coefficients d'échelle

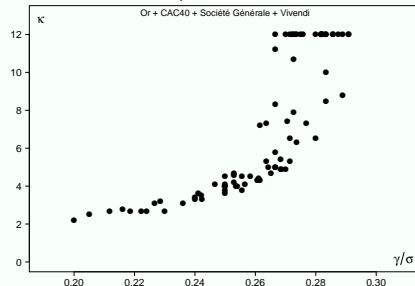


Figure : Comparison of various parameters obtained from  $K1$  distribution (one point = one year  $\approx$  250 days) :  
(a) Median and mean - (b)  $\kappa$  and kurtosis - (c)  $\kappa$  and ratio  $\gamma/\sigma$

## Asymmetric power functions - K2

We call *power functions* the positive functions that verify the equation:

### Power functions

$$\left( y^{1/\alpha} + \frac{X - \mu}{\kappa\gamma} \right) y^{1/\omega} = 1 \quad (4)$$

where  $\kappa$  is the harmonic mean of  $\alpha$  and  $\omega$  such that  $\frac{1}{\kappa} = \frac{1}{2} \left( \frac{1}{\alpha} + \frac{1}{\omega} \right)$

We call *distribution K2* the cumulative function  $F$  and the density function  $f$  defined by:

$$p = F(X) = \frac{1}{1+y} \quad \text{et} \quad f(X) = \frac{dF(X)}{dX} \quad (5)$$

$X$  has a simple form in  $y$  and  $p$ . since  $y = \frac{1-p}{p}$  and  $\log(y) = -\text{logit}(p)$ , it comes:

$$X = \mu + \gamma\kappa \left( -y^{1/\alpha} + y^{-1/\omega} \right) = \mu + \gamma\kappa \left( -\left( \frac{p}{1-p} \right)^{-1/\alpha} + \left( \frac{p}{1-p} \right)^{1/\omega} \right)$$

### Quantile of K2 distribution

$$X(p; \mu, \gamma, \alpha, \omega) = \mu + \gamma\kappa \left( -e^{\frac{-\text{logit}(p)}{\alpha}} + e^{\frac{\text{logit}(p)}{\omega}} \right) \quad (6)$$

## K3 and K4 models : another form for K2 and an extension to K1

$\alpha$  and  $\omega$  are naturally highly correlated within model K2. Consider the terms  $\epsilon$  and  $\delta$ :

$$\frac{1}{\kappa} = \frac{1}{2} \left( \frac{1}{\alpha} + \frac{1}{\omega} \right) \quad \text{et} \quad \delta = \frac{\epsilon}{\kappa} = \frac{1}{2} \left( -\frac{1}{\alpha} + \frac{1}{\omega} \right)$$

It comes:

Conversion from  $\alpha$  and  $\omega$  to and from  $\kappa$ ,  $\delta$  and  $\epsilon$

$$\frac{1}{\alpha} = \frac{1}{\kappa} - \delta \quad \frac{1}{\omega} = \frac{1}{\kappa} + \delta \quad \epsilon = \frac{\alpha - \omega}{\alpha + \omega} \quad \alpha = \frac{\kappa}{1 - \epsilon} \quad \omega = \frac{\kappa}{1 + \epsilon} \quad (7)$$

$-1 < \epsilon < 1$  is a measure of the model **excentricity**. It can be expressed as a %  
 $-\frac{1}{\kappa} < \delta < \frac{1}{\kappa}$  is a measure of the model **distorsion**. It can be expressed in % or ‰

Rewrite the model K2:

Quantiles for models K3 and K4

$$\begin{cases} X(p; \mu, \gamma, \kappa, \delta) = \mu + 2\gamma\kappa \sinh\left(\frac{\text{logit}(p)}{\kappa}\right) e^{\delta \text{logit}(p)} \\ X(p; \mu, \gamma, \kappa, \epsilon) = \mu + 2\gamma\kappa \sinh\left(\frac{\text{logit}(p)}{\kappa}\right) e^{\frac{\epsilon}{\kappa} \text{logit}(p)} \end{cases} \quad (8)$$

Remark:  $e^{\delta \text{logit}(p)} = \left(\frac{p}{1-p}\right)^{\delta}$

# Asymmetric models (K2, K3, K4)

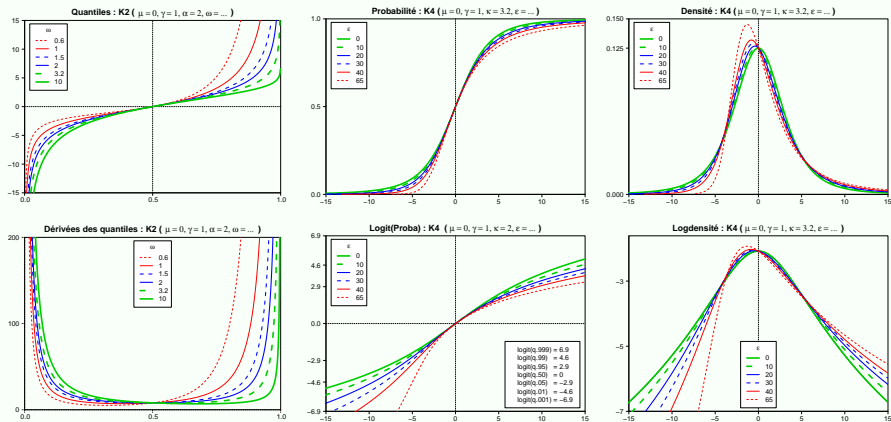


Figure : Models K2 and K3 : (a) Quantiles - (b) Cumulative functions (c) Densities (d) Quantiles derivatives - (e) Logit of the cumulative functions - (f) Logdensities



# Comparison with noncentral Student distribution

Noncentral Student distribution with  $\nu$  degrees of freedom and noncentral parameter  $\mu$  :

$$F_{\nu, \mu}(x) = \begin{cases} \frac{1}{2} \sum_{j=0}^{\infty} \frac{1}{j!} (-\mu\sqrt{2})^j e^{-\frac{\mu^2}{2}} \frac{\Gamma(\frac{j+1}{2})}{\Gamma(1/2)} I\left(\frac{\nu}{\nu+x^2}; \frac{\nu}{2}, \frac{j+1}{2}\right), & x \geq 0 \\ 1 - \frac{1}{2} \sum_{j=0}^{\infty} \frac{1}{j!} (-\mu\sqrt{2})^j e^{-\frac{\mu^2}{2}} \frac{\Gamma(\frac{j+1}{2})}{\Gamma(1/2)} I\left(\frac{\nu}{\nu+x^2}; \frac{\nu}{2}, \frac{j+1}{2}\right), & x < 0 \end{cases}$$

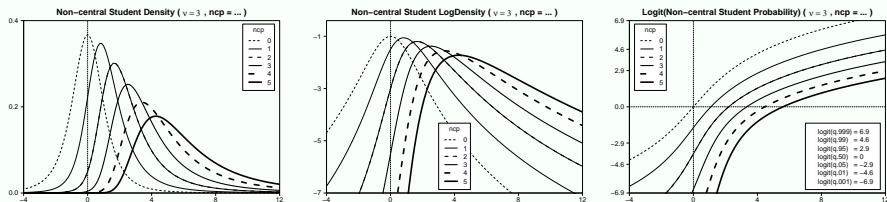


Figure : Noncentral Student distribution ( $\nu = 3$ ,  $ncp = 0, 1, 2, 4, 5$ )  
(a) Densities - (b) Logdensities - (c) Logit of the cumulative functions)

- The noncentral parameter  $\mu$  can take any real value
- It **simultaneously** modifies the distribution tails AND the mode, the median and the mean

# Asymmetric processes ( $\kappa = 2, 3.2, 5, 10$ ) $\times(\delta = -0.04, -0.08, -0.12)$

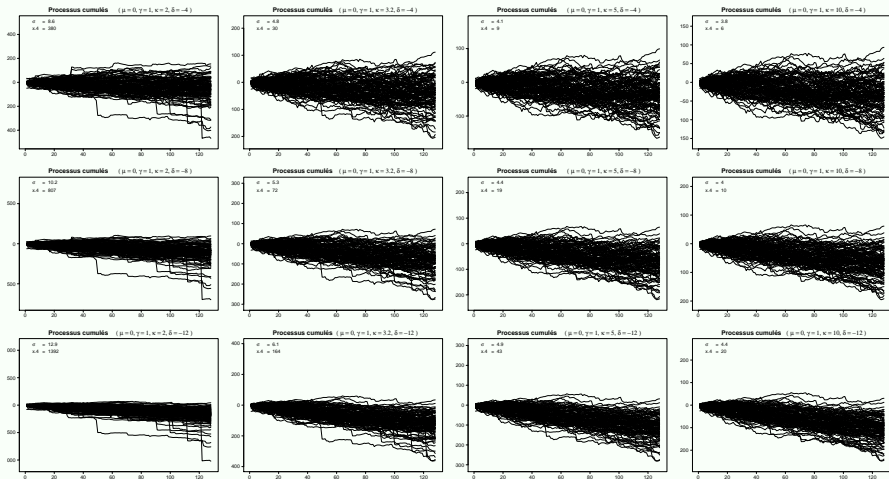
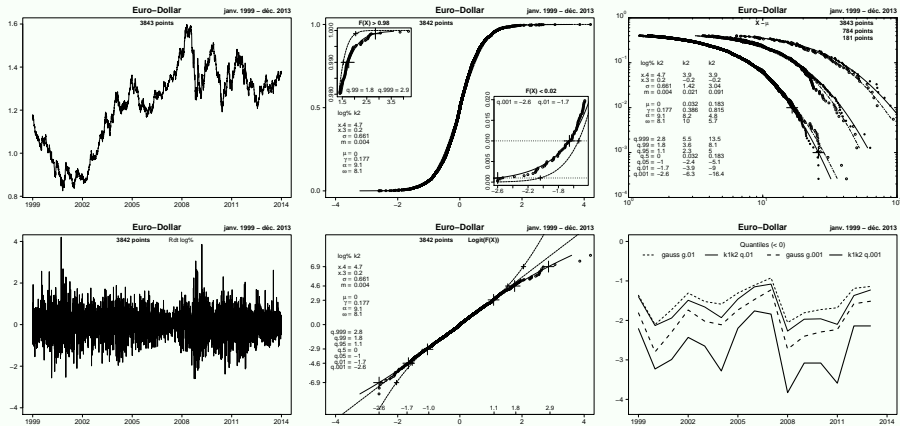


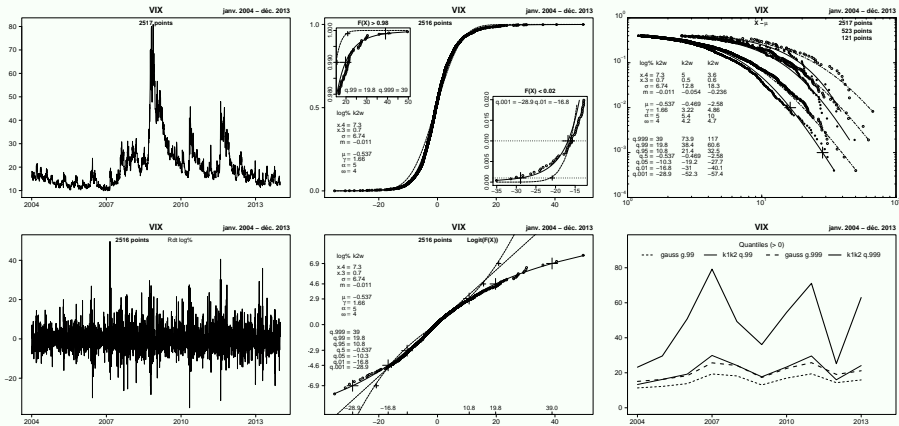
Figure : Cumulative processes: (a)  $\kappa = 2$  (b)  $\kappa = 3.2$  (c)  $\kappa = 5$  (d)  $\kappa = 10$

# Euro-Dollar: $\alpha = 9.1$ , $\omega = 8.1$ , $\epsilon = 6\%$ , multi-scale asymmetry



**Figure :** (a) Euro-Dollar price - (b) 100xlogreturns of Euro-Dollar  
 (c) Cumulative function ( $\alpha > \kappa > \omega$ ) - (d) Logit of the cumulative function  
 (e) Cumulative function in Log-Log scale with periods day, week, month  
 (f) Risks at 1% et 10% over a yearly period (250 days) described by Laplace-Gauss and K2 estimates

# VIX: $\alpha = 5$ , $\omega = 4$ , $\epsilon = 11\%$ , strong multi-scale asymmetry



**Figure :** (a) VIX price - (b) 100xlogreturns of VIX  
 (c) Cumulative function ( $\alpha > \kappa > \omega$ ) - (d) Logit of the cumulative function  
 (e) Cumulative function in Log-Log scale with periods day, week, month  
 (f) Risks at 1% at 1‰ over a yearly period (250 days) described by Laplace-Gauss and K2 estimates

## Miscellaneous

- `ashp`, `kashp(x, k = 1)`, `dkashp_dx`, `invlogit(x)`, `logit(p)`

## Power hyperbolas, power hyperbolic functions and their inverses

- `exphp(x, k = 1)`, `coshp`, `sinhp`, `tanhp`, `sechp`, `cosechp`, `cotanhp`
- `loghp(x, k = 1)`, `acoshp`, `asinhp`, `atanhp`, `asechp`, `acosechp`, `acotanhp`

## Logishp and Kiener1 symmetric functions, without or with parameter m and g

- d,p,q,r `logishp(xqpn, k = 1)`
- d,p,q,r `kiener1(xqpn, m = 0, g = 1, k = 3.2)`

## Kiener2, Kiener3, Kiener4 asymmetric functions

- q,r `kiener2(pn, m = 0, g = 1, a = 3.2, w = 3.2)`
- q,r `kiener3(pn, m = 0, g = 1, k = 3.2, d = 0)`
- q,r `kiener4(pn, m = 0, g = 1, k = 3.2, e = 0)`

## Parameter estimation

- `laplacegaussnorm(X)`
- `regkienerLX(X, model = "k4", dgts = c(3, 3, 1, 1, 1, 3, 2, 4, 4, 2, 2), maxk = 10, mink = 0.7, app = 0)`

⇒ Available on CRAN at:

<http://cran.r-project.org/web/packages/FatTailsR/index.html>

## Package FatTailsR: *regkienerLX* and SMI indice

```
> library(FatTailsR)
> price2returns <- function(x) { 100*diff(log(x)) }
> j <- 2 # 1=DAX, 2=SMI, 3=CAC, 4=FTSE
> X <- price2returns(EuStockMarkets[,2])
> reg <- regkienerLX( X, model = "k2" )
> attributes(reg)
$names
 [1] "dfrXP" "dfrXL" "dfrXR" "dfrEP" "dfrEL" "dfrED" "regk0" "coefk0" "vcovk0"
[10] "vcovk0m" "mcork0" "coefk" "coefk1" "coefk2" "coefk3" "coefk4" "quantk" "coefr"
[19] "coefr1" "coefr2" "coefr3" "coefr4" "quantr" "dfrQkPk" "dfrQkLk"

$class
[1] "clregk"

> reg$coefr2
      m      g      a      w
0.089 0.223 3.800 4.400

> reg$coefr
      m      g      a      k      w      d      e
0.089 0.223 3.800 4.100 4.400 -0.018 -0.070

> reg$quantr
q.0001 q.0005 q.001 q.005 q.01 q.05 q.50 q.95 q.99 q.995 q.999 q.9995 q.9999
-10.00 -6.44 -5.30 -3.29 -2.63 -1.42 0.09 1.44 2.39 2.88 4.29 5.06 7.33
```

We have presented:

- Power hyperbolas, power hyperbolic functions and fat-tailed distributions which use the **median** as pivotal value
  - A symmetric model whose all representations have explicit forms (pdf, cdf, quantiles, ...)
  - A few asymmetric models whose quantile functions have explicit forms
- Very accurate models for distributions of returns. Accurate estimates of risks
- A new R package: **FatTailsR**

Some open questions:

- **Characteristic functions**, moments, Bayesian prior and posterior
- Multivariate models, copules. But which **scalar product** and which **correlation** ?  
Correlation of central parameter  $\gamma$ , tail parameter  $\kappa$  or quantile value  $q_{1\%}$  ?
- **Stochastic processes**. Here,  $\mu\gamma\kappa\epsilon$  are not  $m\sigma$  !!  
What name for these new processes:  $\mu\gamma\kappa\epsilon$ -Garch or maybe Garck ?

$$\frac{dS_t}{S_t} = \mu_m dt + \gamma dW_t \quad \text{with} \quad dW_t \approx \approx 2\kappa \sinh\left(\frac{\text{logit}(p_t)}{\kappa}\right) e^{\frac{\epsilon}{\kappa} \text{logit}(p_t)}$$

- Could the nonlinear variation of parameters ( $\gamma/\sigma$ ,  $\kappa$ ) presented at slide 20 explain the **volatility smile** of derivatives products?

Potential applications:

- Market risk (Bâle III, bcbs240, Jan-Fev. 2013), portfolio management, derivatives, ...
- Add new features to the package

- 1 B. Mandelbrot, Sur certains prix spéculatifs : faits empiriques et modèle basé sur les processus stables additifs non gaussiens de Paul Lévy, C. R. Acad. Sci. Paris, vol. 254, (1962) 3968-3970 (reprinted in Fractales, hasard et finance, Flammarion, 2009).
- 2 J-P. Bouchaud, M. Potters, Théorie des risques financiers, CEA Aléa Saclay, 1997.
- 3 L. De Haan, On regular variation and its application to the weak convergence of sample extremes, Mathematical Centre Tracts vol. 32, Mathematisch Centrum Amsterdam, 1970.
- 4 The R Project for Statistical Computing, <http://www.r-project.org>
- 5 Rmetrics, <http://www.rmetrics.org>
- 6 8th R/Rmetrics Workshop and Summer School, <https://sites.google.com/site/rmetricsparis2014/>

## Package **FatTailsR**

- Version 1.0-3 (14 July 2014) available on CRAN at:  
<http://cran.r-project.org/web/packages/FatTailsR/index.html>

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<http://www.inodelia.com/fattailsr-en.html>





**Thank you for your attention !**

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