

III. Short rate models: Evolution of the short rate

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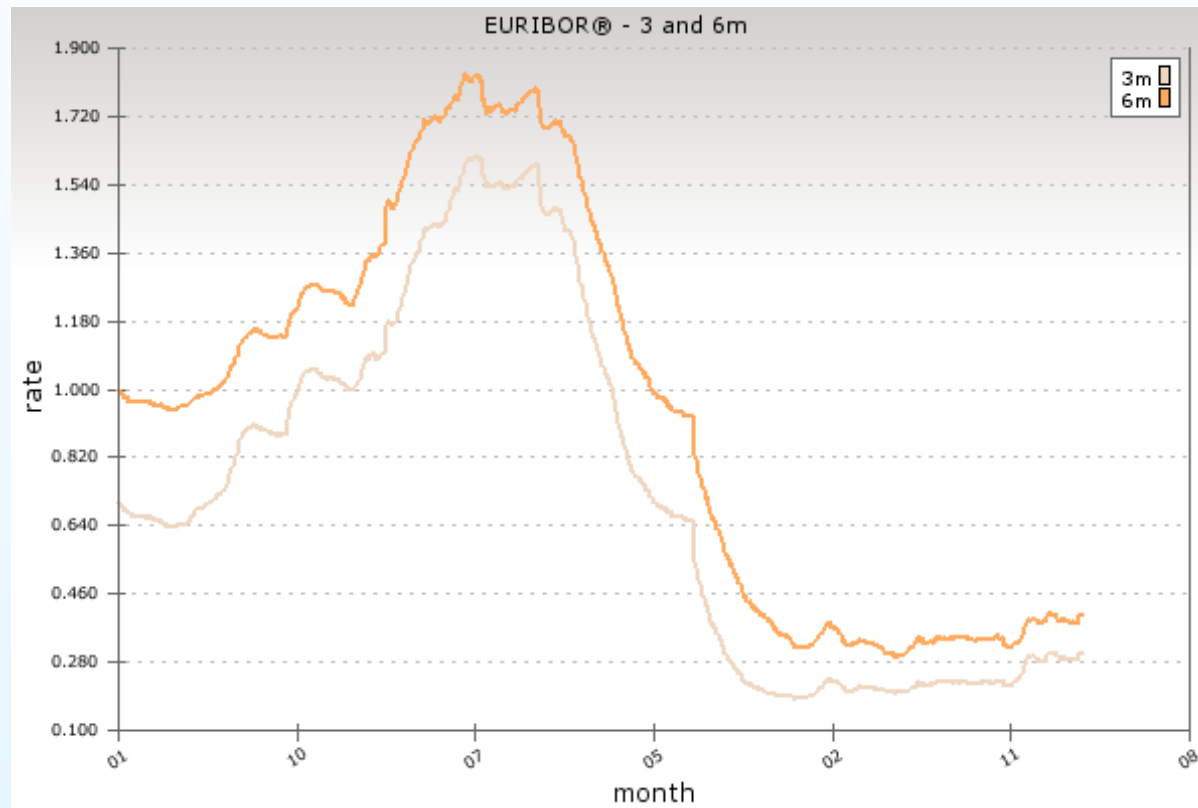
Comenius University, Bratislava

Interest rate

- Assumptions in several models: the interest rate is constant , for example when pricing a riskless portfolio in the derivation of the Black-Scholes model ($dP = rPdt$, cf. *partial differential equations* lectures)
- Reasonable in some cases, but not - for example - if the derivative directly depends on the interest rate (bond, swap, ...)
- What we need to model:
 - the interest rate is not constant
 - there are interest rates with different maturities

Interest rates

- Example: Euribor (European Inter-Bank Offered Rate):



<http://www.euribor-ebf.eu/>

Short rate models

- Short rate - it is the instantaneous interest rate - interest rate for an infinitesimally small time interval
- Theoretical variable, in practice we use a proxy (1 month, 3 months)
- Short rate models:
 - Short rate r is modelled by a stochastic differential equation

$$dr = \mu(r, t)dt + \sigma(r, t)dw$$

Terminology: $\mu(r, t)$ - drift, $\sigma(r, t)$ - volatility

- Other interest rates and derivatives - solving a partial differential equation

Mean-reversion models

- Mean-reversion - reverting to some long-term equilibrium level
- This property in short rate models: the drift is taken to be

$$\mu(r, t) = \kappa(\theta - r)dt,$$

where $\kappa, \theta > 0$ are constants

- ODE for the expected value $\mathbb{E}[r]$ (for a given r_0):

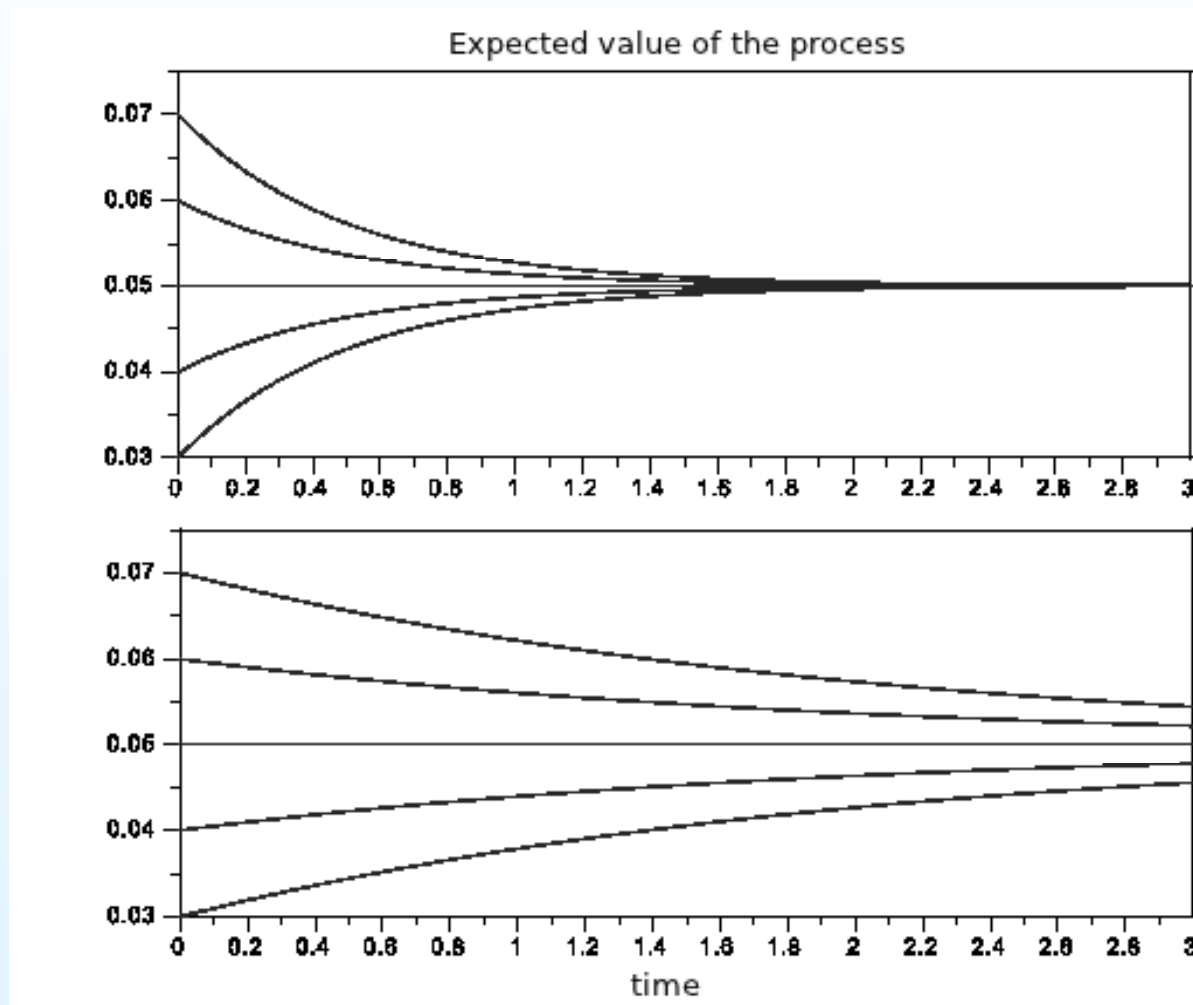
$$d\mathbb{E}[r] = \kappa(\theta - \mathbb{E}[r])dt + \mathbb{E}[\sigma(r, t)dw] = \kappa(\theta - \mathbb{E}[r])dt,$$

its solution is: $\mathbb{E}[r_t] = r_0e^{-\kappa t} + (1 - e^{-\kappa t})\theta$

- Therefore $\mathbb{E}[r] \rightarrow \theta$ as $t \rightarrow \infty$

Mean-reversion models

- Sample solutions for selected r_0 :



What is θ ? What parameter is different in these two cases?

Mean-reversion models

- Example: $dr = \kappa(\theta - r)dt + \sigma dw$ Ornstein-Uhlenbeck process, in finance known as Vasicek model



- Oldřich Alfons Vašíček (born 1942) - a Czech mathematician
- Emigrated to the USA in 1968
- 1969: employed in the management science department of Wells Fargo Bank.

Photo: <http://www.risk.net/risk-magazine/feature/1506410/presenting-risk-awards-2002>

About Vasicek: <http://www.risk.net/risk-magazine/feature/1506624/2002-winner-lifetime-achievement-award-oldrich-alfons-vasicek>

Examples of one-factor models

- Already mentioned Vasicek model
 - Short rate: $dr = \kappa(\theta - r)dt + \sigma dw$
 - Drawback: allows negative interest rates (intuitively for now: also for r very close to zero, the volatility is always the same)
- Cox-Ingersoll-Ross:
 - J. C. Cox., J. E. Ingersoll Jr, S. A. Ross, **A theory of the term structure of interest rates**, *Econometrica* (1985) 385-407.
 - Short rate: $dr = \kappa(\theta - r)dt + \sigma\sqrt{r}dw$
 - Solves the previous problem: does not allow negative interest rates (intuitively: for $r = 0$, the volatility is zero and the drift is positive)
 - It can be shown that if $2\kappa > \sigma^2$, then $r = 0$ has a zero probability (intuition: SDE for $y = \ln(r)$ and analysis of the drift)

Examples of one-factor models

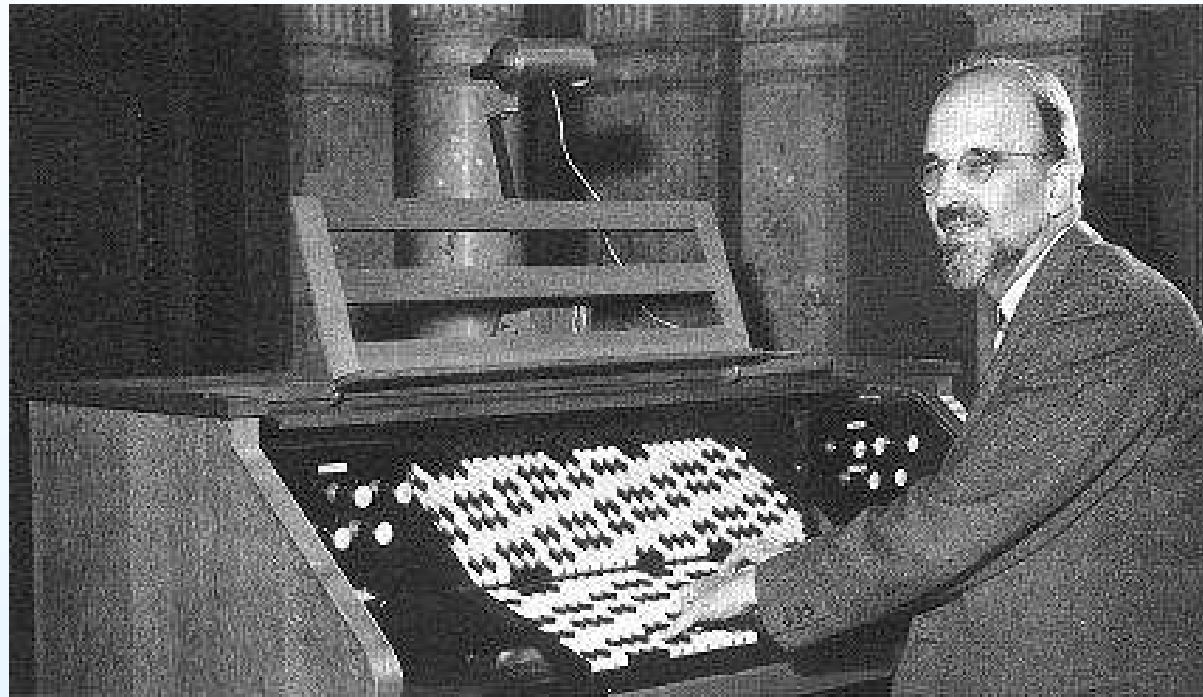
- Chan-Karolyi-Longstaff-Sanders:
 - C. K. Chan, G. A. Karolyi, F. A. Longstaff, A. B. Sanders, **An empirical comparison of alternative models of the short-term interest rate**, The Journal of Finance 47 (1992) 1209-1227.
 - Short rate: $dr = \kappa(\theta - r)dt + \sigma r^\gamma$
 - Vašíček a CIR are special cases ($\gamma = 0, \gamma = 1/2$)
 - They estimated a general model (optimal γ turned out to be 1.5) and tested $\gamma = 0, \gamma = 1/2$ as restrictions on parameters → they were rejected
 - Later many other studies of this kind (different data sets, different statistical methods)

Fokker-Planck PDE

- Fokker-Planck PDE - partial differential equation for the density of probability distribution of the value of a stochastic process
- Out of curiosity:
 - Max Karl Ernst Ludwig Planck (1858-1947) was singing, playing the piano, organ and cello, composed songs and opera, ... but he decided to study physics
 - Adriaan Daniël Fokker (1887-1972) was interested in microtonal music, proposed a 31-tonal organ which was exhibited in *Teylers Museum* v Haarleme (the oldest museum in the Netherlands, Fokker was a curator of the physical cabinet)

Fokker-Planck PDE

A. D. Fokker and his organ:



This and other photos: <http://www.huygens-fokker.org/instruments/fokkerorgan.html>

Fokker-Planck PDE

- Consider the following process

$$dx = \mu(x, t)dt + \sigma(x, t)dw$$

and define $g(x, t)$ as a conditional density of the value of the process at time t if the value x_0 at time $t = 0$ is given

- THEOREM:

Then the function $g(x, t)$ is a solution to the Fokker-Planck PDE

$$\frac{\partial g}{\partial t} = \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 g) - \frac{\partial}{\partial x} (\mu g)$$

with initial condition $g(x, 0) = \delta(x - x_0)$.

Fokker-Planck PDE

Remark on function δ from the initial condition - it is so called Dirac function:

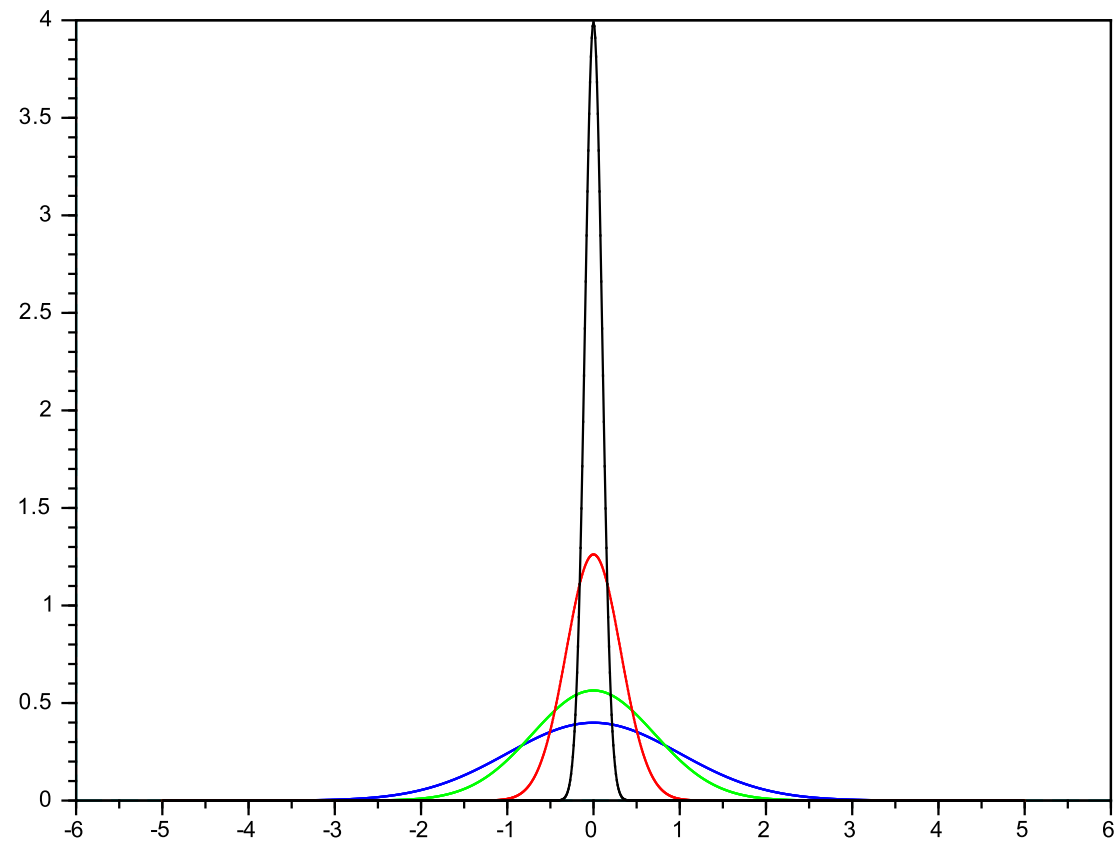
- Not a function in the classical sense
- Intuition:
 - function satisfying

$$\delta(x-x_0) = \begin{cases} 0 & \text{for } x \neq x_0 \\ +\infty & \text{for } x = x_0 \end{cases}, \quad \int_{-\infty}^{\infty} \delta(x-x_0) dx = 1.$$

- "density" of a random variable which takes the value x_0 with probability 1
- We have: $\int_{-\infty}^{\infty} \delta(x-x_0) f(x) dx = f(x_0)$
- It can be defined in a mathematically precise way (we will not do this)

Fokker-Planck PDE

Intuitively - functions "converging" to a Dirac function:



Fokker-Planck PDE: proof

- Let $V = V(x, t)$ be an arbitrary function with compact support, i.e., $V \in C_0^\infty(\mathbb{R} \times (0, T))$
- From Itô lemma:

$$dV = \left(\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} \frac{\partial^2 V}{\partial x^2} + \mu \frac{\partial V}{\partial x} \right) dt + \sigma \frac{\partial V}{\partial x} dW.$$

- Let E_t is the expected value with respect to the distribution given by the density $g(x, t)$
- Then

$$dE_t(V) = E_t(dV) = E_t \left[\left(\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} \frac{\partial^2 V}{\partial x^2} + \mu \frac{\partial V}{\partial x} \right) dt \right].$$

Fokker-Planck PDE: proof

- We have $V(x, 0) = V(x, T) = 0$ and $V(x, t) = 0$ for $|x| > R$ where $R > 0$ is sufficiently large
- Integration *per partes*:

$$\begin{aligned} 0 &= \int_0^T \frac{d}{dt} E_t(V) dt = \int_0^T E_t \left(\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} \frac{\partial^2 V}{\partial x^2} + \mu \frac{\partial V}{\partial x} \right) dt \\ &= \int_0^T \int_{\mathbb{R}} \left(\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} \frac{\partial^2 V}{\partial x^2} + \mu \frac{\partial V}{\partial x} \right) g(x, t) dx dt \\ &= \int_0^T \int_{\mathbb{R}} V(x, t) \left(-\frac{\partial g}{\partial t} + \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 g) - \frac{\partial}{\partial x} (\mu g) \right) dx dt. \end{aligned}$$

- Since $V \in C_0^\infty(\mathbb{R} \times (0, T))$ was arbitrary, for the density $g = g(x, t)$ we obtain the Fokker-Planck equation

Fokker-Planck PDE for the Vasicek model

Let x_t be an Ornstein-Uhlenbeck/Vasicek process

- Constant at $dw \rightarrow$ we can expect normal distribution
- We have already computed the expected values
- We derive an equation for the variance:

```
(%i1) mi(t):=x0*exp(-kappa*t)+theta*(1-exp(-kappa*t)) $  
(%i2) g(x,t):=exp(-(x-mi(t))^2/(2*s2(t)))/sqrt(2*pi*s2(t));  
(%o2) g(x,t):=
$$\frac{\exp\left(\frac{-(x-mi(t))^2}{2s2(t)}\right)}{\sqrt{2\pi s2(t)}}$$
  
(%i3) -diff(g(x,t),t)+  
(sigma^2/2)*diff(g(x,t),x,2)+  
kappa*g(x,t)-  
kappa*(theta-x)*diff(g(x,t),x) $
```

(Computations in the wxMaxima software: <http://wxmaxima.sourceforge.net>)

Fokker-Planck PDE for the Vasicek model

- Factoring the expression:

$$\begin{aligned} & \text{factor(\%);} \\ (\%04) & - \left(\frac{d}{dt} s_2(t) + 2\kappa s_2(t) - \sigma^2 \right) \left(x\theta^2 - 2\%e^{\kappa t} x x\theta + 2\%e^{\kappa t} \theta x\theta - 2\theta x\theta + \%e^{2\kappa t} x^2 - 2\%e^{2\kappa t} \theta x \right. \\ & + 2\%e^{\kappa t} \theta x + \%e^{2\kappa t} \theta^2 - 2\%e^{\kappa t} \theta^2 + \theta^2 - \%e^{2\kappa t} s_2(t) \left. \right) \\ & \frac{\%e^{-2\kappa t} x\theta^2}{2s_2(t)} + \frac{\%e^{-\kappa t} x x\theta}{s_2(t)} - \frac{\%e^{-\kappa t} \theta x\theta}{s_2(t)} + \frac{\%e^{-2\kappa t} \theta x\theta}{s_2(t)} - \frac{x^2}{2s_2(t)} - \frac{\%e^{-\kappa t} \theta x}{s_2(t)} + \frac{\theta x}{s_2(t)} + \frac{\%e^{-\kappa t} \theta^2}{s_2(t)} - \frac{\%e^{-2\kappa t} \theta^2}{2s_2(t)} - \frac{\theta^2}{2s_2(t)} - 2\kappa t \\ & \left. \right) / \\ & (2^{3/2} \sqrt{\pi} s_2(t)^{5/2}) \end{aligned}$$

- More clearly:

$$\begin{aligned} & \text{factor(\%);} \\ (\%04) & - \left(\frac{d}{dt} s_2(t) + 2\kappa s_2(t) - \sigma^2 \right) \left(x\theta^2 - 2\%e^{\kappa t} x x\theta + 2\%e^{\kappa t} \theta x\theta - 2\theta x\theta + \%e^{2\kappa t} x^2 - 2\%e^{2\kappa t} \theta x \right. \\ & + 2\%e^{\kappa t} \theta x + \%e^{2\kappa t} \theta^2 - 2\%e^{\kappa t} \theta^2 + \theta^2 - \%e^{2\kappa t} s_2(t) \left. \right) \\ & \frac{\%e^{-2\kappa t} x\theta^2}{2s_2(t)} + \frac{\%e^{-\kappa t} x x\theta}{s_2(t)} - \frac{\%e^{-\kappa t} \theta x\theta}{s_2(t)} + \frac{\%e^{-2\kappa t} \theta x\theta}{s_2(t)} - \frac{x^2}{2s_2(t)} - \frac{\%e^{-\kappa t} \theta x}{s_2(t)} + \frac{\theta x}{s_2(t)} + \frac{\%e^{-\kappa t} \theta^2}{s_2(t)} - \frac{\%e^{-2\kappa t} \theta^2}{2s_2(t)} - \frac{\theta^2}{2s_2(t)} - 2\kappa t \\ & \left. \right) / \\ & (2^{3/2} \sqrt{\pi} s_2(t)^{5/2}) \end{aligned}$$

- This has to equal to zero

Fokker-Planck PDE for the Vasicek model

- This expression has to be equal to zero \Rightarrow this holds if

$$s2'(t) + 2\kappa s2(t) - \sigma^2 = 0$$

- Variance at time $t = 0$ is zero \Rightarrow initial condition $s2(0) = 0$
- Solution:

$$s2(t) = \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa t})$$

- CONCLUSION:
Distribution of an Ornstein-Uhlenbeck process is a normal distribution with expected value $r_0 e^{-\kappa t} + (1 - e^{-\kappa t})\theta$ and variance $s2(t) = \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa t})$

Vasicek model: estimating parameters

- We have a time serie of interest rate (proxy for the short rate) → we want to estimate the parameters of the Vasicek model
- Knowledge of the conditional distribution allows us to construct the likelihood function for the given values of the interest rate r_1, r_2, \dots, r_n observed in the market:

$$L = \prod_{i=1}^{n-1} f(r_{i+1}|r_i)$$

- Maximizing L (equivalently, its logarithm) yields estimates of the parameters
- Vasicek model: functions f are normal distribution densities; it is possible to find closed form expressions for the estimates; we will use then on exercises session

CKLS model: estimating parameters

- Recall that $dr = \kappa(\theta - r)dt + \sigma r^\gamma$
- Conditional distribution known only for $\gamma = 1/2$; even that quite complicated \rightarrow approximation of the likelihood:
 - The volatility σr^γ on time interval $[t, t + \Delta t)$ between two observations is approximated by its value in time t
 - In this approximation, volatility on $[t, t + \Delta t)$ is constant \rightarrow normal distribution
 - Known as Nowman's Gaussian estimates (since based on Gaussian approximation)
- Maximum likelihood estimates \rightarrow testing hypotheses using likelihood ratio test

CKLS model: estimating parameters - example

Athanasios Episcopos: **Further evidence on alternative continuous time models of the shortterm interest rate**, Journal of International Financial Markets, Institutions and Money Volume 10, Issue 2, June 2000, pp. 199-212

- Estimates for 10 countries (general CKLS model and several restrictions given by existing models)
- Next page: example for the USA ((data: 1/1986 - 4/1998, 148 observations)
- We go through the procedure of testing Vasicek and CIR model as restrictions of the CKLS model (computation of the test statistics and corresponding p-value)

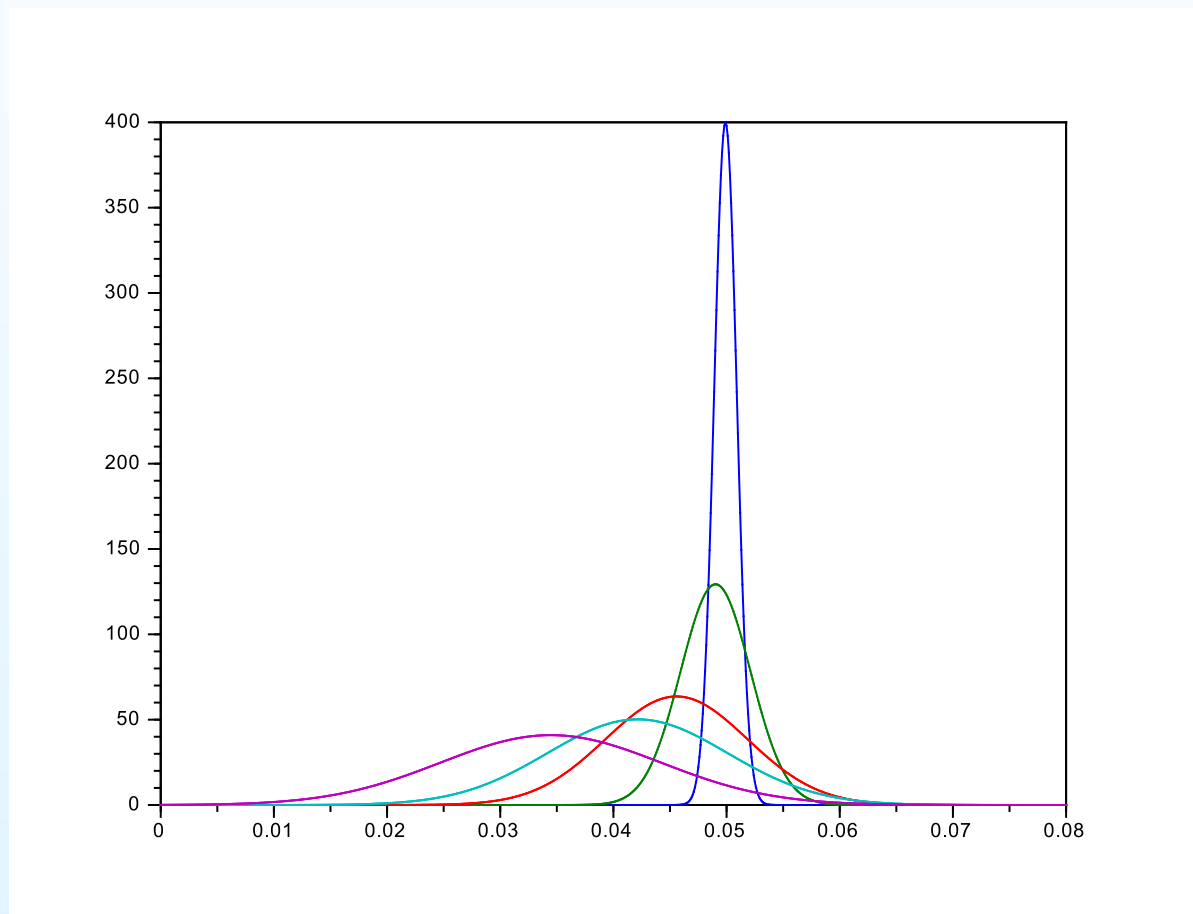
CKLS model: estimating parameters - example

Results for the USA:

Model ^b	α	β	σ^2	γ	Avg. Log L	χ^2 -test ^c	df
Unrestricted	0.0013 (1.4696)	-0.0234 (-1.5710)	0.0001 (1.0309)	0.4239 (2.5099)	5.1768		
Vasicek	0.0013 (4.3499)	-0.0235 (-5.6112)	0.0000 (16.1077)	0	5.1569	5.8655 (0.0154)	1
CIR SR	0.0013 (4.6916)	-0.0241 (-5.8893)	0.0002 (13.0558)	0.5	5.1761	0.201 (0.6539)	1
BR-SC	0.0014 (6.1214)	-0.0255 (-6.1221)	0.0038 (14.9738)	1	5.1365	11.8748 (0.0006)	1
CIR VR	0	0	0.0794 (21.0933)	1.5	5.0220	45.529 (0.0000)	3
CEV	0	-0.003 (-0.6401)	0.0001 (24.7981)	0.4063 (31.0657)	5.1705	1.8477 (0.1740)	1

Fokker-Planck PDE: limiting distribution

- EXAMPLE: Ornstein-Uhlenbeck process - densities for a given x_0 and a couple of times t :



Fokker-Planck PDE: limiting distribution

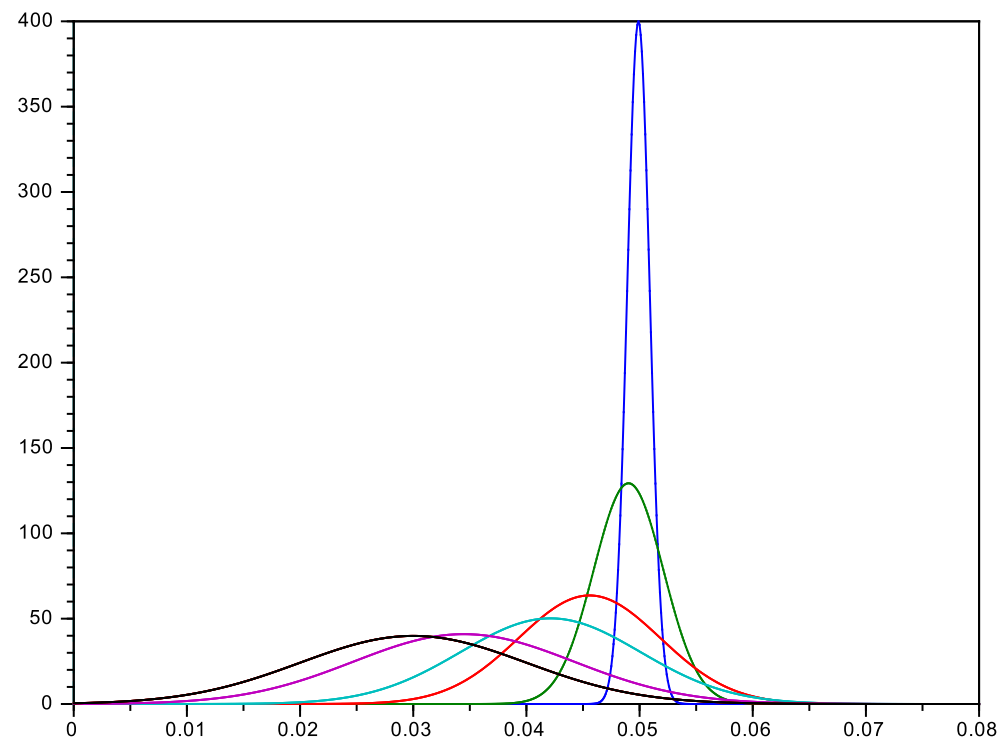
- Densities converge to a certain limiting distribution
- As $t \rightarrow \infty$, we have:

$$\mathbb{E}[r_t|r_0] = r_0e^{-\kappa t} + (1 - e^{-\kappa t})\theta \rightarrow \theta$$

$$\mathbb{D}[r_t|r_0] = \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa t}) \rightarrow \frac{\sigma^2}{2\kappa}$$

Fokker-Planck PDE: limiting distribution

- EXAMPLE: Ornstein-Uhlenbeck process - densities for a given x_0 and a couple of times t (from the previous plot) and the limiting density (black line):



Fokker-Planck PDE: limiting distribution

- We do not need the conditional distributions to compute the limiting distribution (this is often complicated)
- Direct computation from the Fokker-Planck PDE:
 - we know that the density $g(x, t)$ for time t satisfies the PDE

$$\frac{\partial g}{\partial t} = \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 g) - \frac{\partial}{\partial x} (\mu g)$$

- consider limit $f(x) := \lim_{t \rightarrow \infty} g(x, t)$
- this limit then satisfies the stationary Fokker-Planck equation:

$$0 = \frac{1}{2} \frac{d^2}{dx^2} (\sigma^2 f) - \frac{d}{dx} (\mu f)$$

with a normalization condition (it is a density function)

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

Fokker-Planck PDE: limiting distribution

EXAMPLE: CIR model for interest rates

- Recall the stochastic differential equation for the short rate
 $dx = \kappa(\theta - x)dt + \sigma\sqrt{x}dw$
- Let $2\kappa > \sigma^2$ (then the zero value cannot be achieved)
- Density $g(x, t)$ at time t satisfies the PDE

$$\frac{\partial g}{\partial t} = \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 g) - \frac{\partial}{\partial x} (\kappa(\theta - x)g)$$

with initial condition

$$g(x, 0) = \delta(x - x_0)$$

- There is an explicit expression for $g(x, t)$, but it is quite complicated (noncentral chi-squared distribution, modified Bessel function)

Fokker-Planck PDE: limiting distribution

EXAMPLE - CONTINUED:

- Explicit solution (without a proof):

$$g(x, t) = c e^{-u-v} \left(\frac{v}{u}\right)^{q/2} I_q(2\sqrt{uv}),$$

(for $x > 0$, otherwise $g(x, t) = 0$), where I_q is a modified Bessel function of the first kind of order q and

$$c = \frac{2\kappa}{\sigma^2(1 - e^{-\kappa t})}$$

$$u = c x_0 e^{-\kappa t}$$

$$v = c x$$

- Complicated, but limiting distribution can be found also without a knowledge of this conditional distribution

Fokker-Planck PDE: limiting distribution

EXAMPLE - CONTINUED:

- Limiting density $f(x) := \lim_{t \rightarrow \infty} g(x, t)$ satisfies stationary Fokker-Planck equation: (for $x > 0$, otherwise it is zero since the process never has negative values)

$$0 = \frac{1}{2} \frac{d^2}{dx^2} (\sigma^2 x f) - \frac{d}{dx} (\kappa(\theta - x) f)$$

- HOMEWORK: Integrating gives

$$f(x) = K x^{\frac{2\kappa\theta}{\sigma^2} - 1} e^{-\frac{2\kappa}{\sigma^2} x}$$

- Constant K is computed from the condition $\int_{-\infty}^{\infty} f(x) dx = 1$
- Note that this is a density of a gamma distribution