

# Dynamic Stochastic Accumulation Model

## with Application to Risk Management in Slovak Pension System

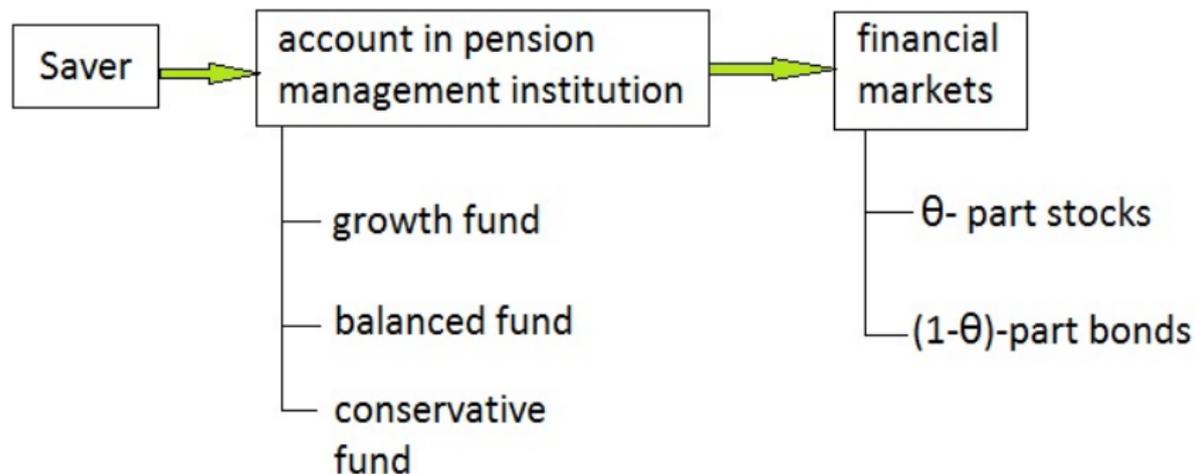
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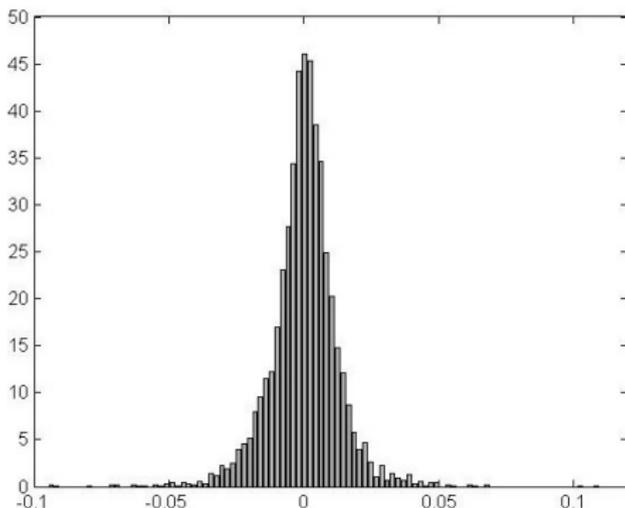
13<sup>th</sup> of May 2012

- Saving problem
- Modeling of returns with fat-tailed distributions
- Accumulation model for II. pillar of Slovak pension system
- Numerical computation
- Results with the NIG distributed returns
- Conclusions

# Pension Problem - II. Pillar

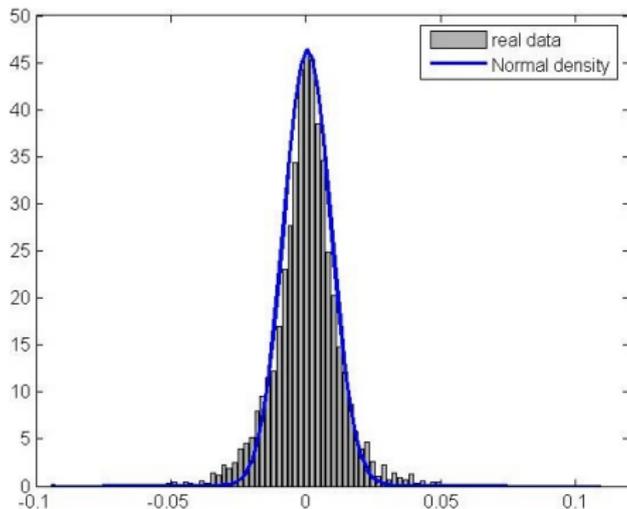


**Figure:** Histogram of log returns for daily data for S&P500 Index (Jan-1996 - Feb-2012).



# Modeling Returns - Normal distribution

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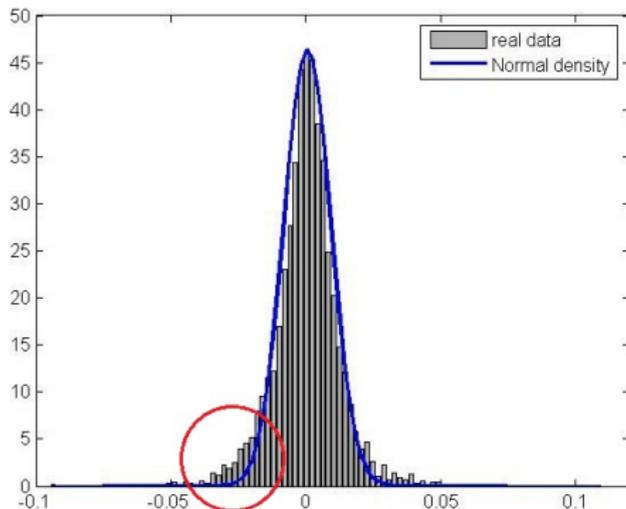


## Parameters:

mean value: 0.02%; standard deviation: 1.31%;

# Modeling Returns - Normal distribution

**Figure:** Histogram of log returns for daily data for S&P500 Index (Jan-1996 - Feb-2012).

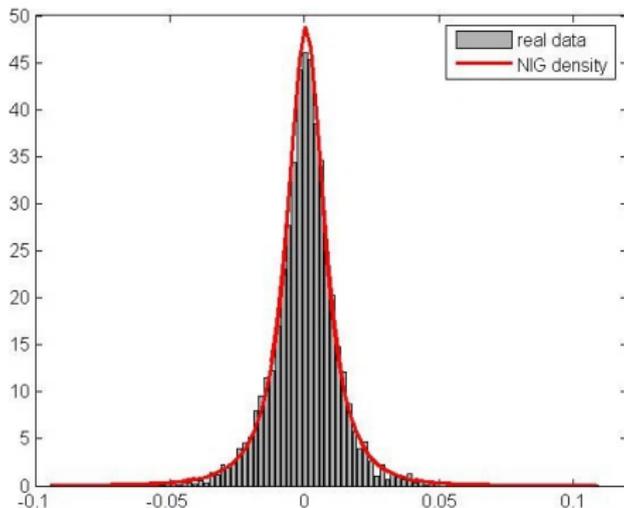


## Parameters:

mean value: 0.02%; standard deviation: 1.31%;

# Modeling Returns - Fat-tailed distribution

**Figure:** Histogram of log returns for daily data for S&P500 Index (Jan-1996 - Feb-2012).



## Parameters:

mean value: 0.02%; standard deviation: 1.31%;

skewness: -0.22; kurtosis: 10.08;

$$\max_{\theta} \mathbb{E}(U(d_T)),$$

subject to

$$\begin{aligned}d_{t+1} &= F_t(d, \theta(d, t), r_t^\theta) \\ d_1 &= \tau\end{aligned}$$

where  $T$  is the retirement time,  $d_t$  is a measure of saved amount of money and  $\theta$  is the selected stock-to-bond proportion with saver's utility function  $U$ .

# Bellman Equation

Using the tower law for conditional expectation, we obtain the Bellman equation

$$V_T(d) = U(d),$$

$$\begin{aligned} V_t(d) &= \max_{\theta} \mathbb{E}(V_{t+1}(d_{t+1})) = \max_{\theta} \mathbb{E}(V_{t+1}(F_t(d, \theta(d, t), r_t^\theta))) \\ &= \max_{\theta} \int_{\mathbb{R}} V_{t+1}(d_{t+1}) f^\theta(r) dr. \end{aligned}$$

The problem we solve in each time step  $t$  is to find the maximum over  $\theta$  of the integral

$$\int_{\mathbb{R}} V_{t+1}(d_{t+1}) f^\theta(r) dr.$$

## Definition

The random variable  $X$  is Normal Inverse Gaussian distributed  $NIG(\alpha, \beta, \mu, \delta)$  if its probability density function is given by

$$f(x) = \frac{\alpha}{\pi} \exp\{\delta\sqrt{\alpha^2 - \beta^2} + \beta(x - \mu)\} \frac{K_1(\alpha\delta\sqrt{1 + (\frac{x-\mu}{\delta})^2})}{\sqrt{1 + (\frac{x-\mu}{\delta})^2}}$$

where  $K_1$  denotes the modified Bessel function of the third kind,  $K_1(x) = \frac{1}{2} \int_0^\infty e^{-\frac{x}{2}(y+y^{-1})} dy$ , and the conditions for the parameters are  $\alpha > 0, \delta > 0, \mu \in \mathbb{R}, 0 \leq |\beta| \leq \alpha$ .

# Properties of NIG Distribution

**The problem:** What will be the distribution of the convolution of two NIG distributed random variables if their parameters will be different?

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The NIG class of densities has the following properties:

- 1** Scaling property: If  $X \sim NIG(\alpha, \beta, \mu, \delta)$ , then  $Y = cX \sim NIG(\alpha/c, \beta/c, c\mu, c\delta)$ .
- 2** Convolution property: If  $X_1 \sim NIG(\alpha, \beta, \mu_1, \delta_1)$  and  $X_2 \sim NIG(\alpha, \beta, \mu_2, \delta_2)$  are independent, then the sum  $Y = X_1 + X_2 \sim NIG(\alpha, \beta, \mu_1 + \mu_2, \delta_1 + \delta_2)$ .
- 3** Standardization: If  $X \sim NIG(\alpha, \beta, \mu, \delta)$ , then variable  $Y = (X - \mu)/\delta$  has the *Standard Normal Inverse Gaussian Distribution*  $NIG(\alpha\delta, \beta\delta, 0, 1)$ .

# Numerical Approximation of Portfolio Parameters

The procedure:

- 1 generation of the random vectors:

$$Z_B \sim NIG(\alpha_B \delta_B, \beta_B \delta_B, 0, 1)$$

$$Z_S \sim NIG(\alpha_S \delta_S, \beta_S \delta_S, 0, 1)$$

- 2 construction of the correlated random vectors so that

$$X_B = \delta_B Z_B + \mu_B$$

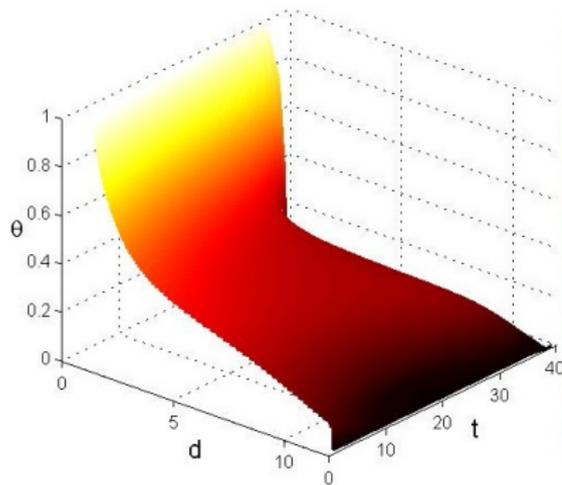
$$X_S = \delta_S (\rho_{B,S} Z_B + \sqrt{1 - \rho_{B,S}^2} Z_S) + \mu_S$$

where the vectors  $X_B$  and  $X_S$  represent the bond and stock log-returns, respectively.

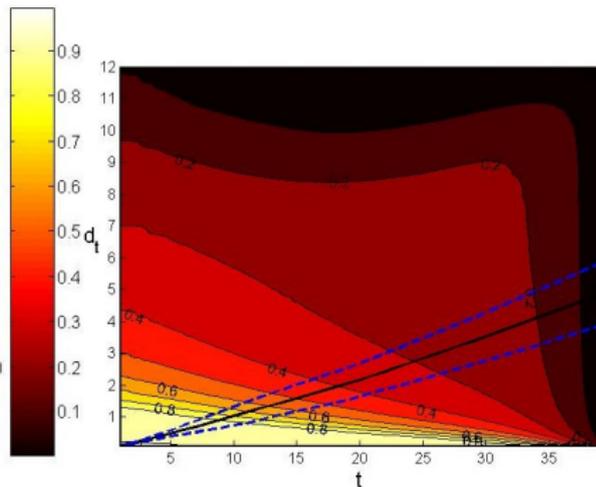
- 3 computing the log-return of the portfolio with stock to bond proportion  $\theta$  in the way:  $X^\theta = \theta X_S + (1 - \theta) X_B$
- 4 computing the values of the first four moments

→ **NIG parameters**

# Optimal Proportion in Portfolio



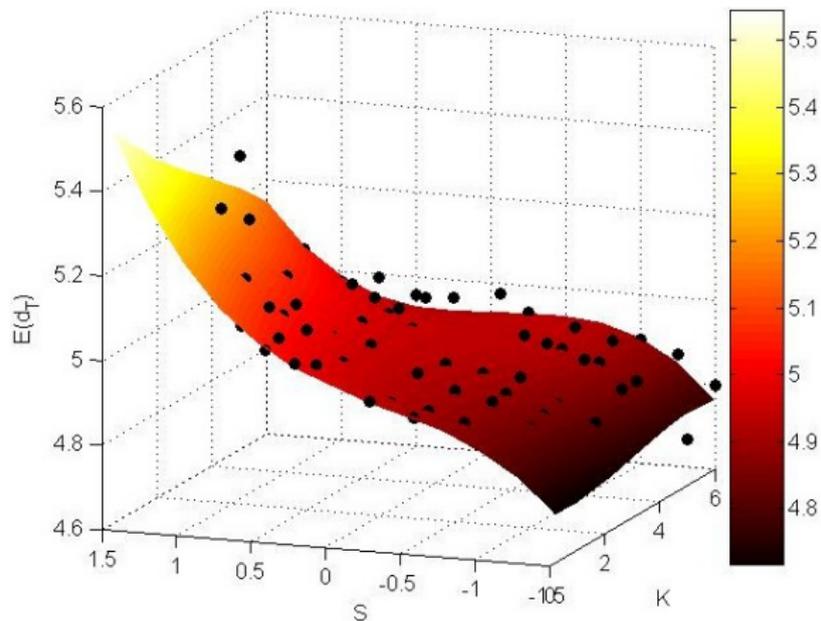
(a) 3D plot of optimal choice of  $\theta$



(b) contour

**Figure:** 3D plot of the of the optimal choice of  $\theta$  as a function of time  $t$  and current accumulated sum  $d_t$  for  $S = -0.2, \mathcal{K} = 10$ .

# Sensitivity Analysis



**Figure:** 3D plot of the final mean value of the accumulated sum as a function of skewness  $S$  and excess kurtosis  $K$ . Figures shows different views.

- We have introduced a saving problem.
- We have analyzed S&P500 Index in more detail and showed that the Index is characterized by higher kurtosis and is skewed to the left.
- We have presented the dynamic stochastic model for determining the optimal value of the stock to bond proportion in pension funds.
- We have formulated the problem in term of Bellman equation.
- We have presented the numerical procedure and results.
- We have shown different levels of skewness and kurtosis of the underlying assets influence the optimal choice and the level of accumulated sum.

Thank you for your attention.