

# **Two-Parameter Analysis of Rician Data: Basics of a Theory and Computer Simulation by Wolfram Mathematica**

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The Rice statistical distribution characterizes a value of an amplitude, or envelope of the complex signal composed as a sum of any determined initial signal and a Gaussian noise.

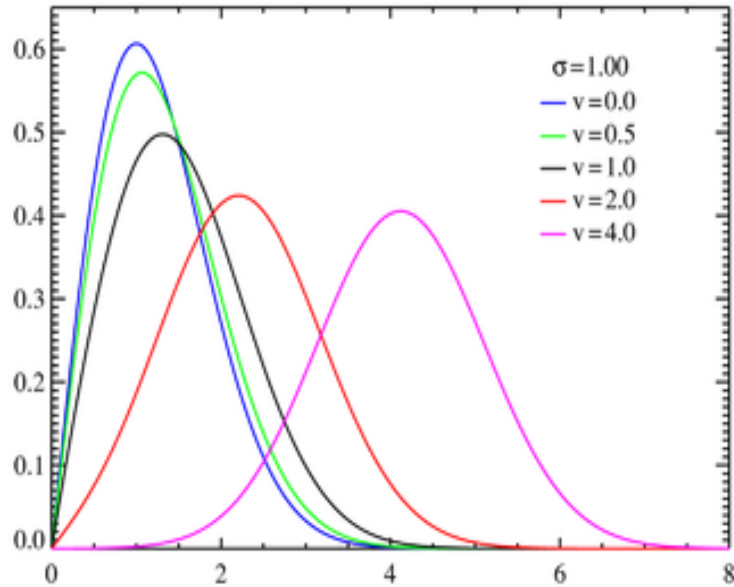
The fields of application:

- *magnetic-resonance visualization,*
- *radio signals reception and processing,*
- *radar signals analysis,*
- *analysis of the sonar signals,*
- *optical medium's properties measurements, etc.*

# Main points of the Speech:

- The applicability of the Rice statistical model at solving many typical tasks of data processing;
- Brief review of the problem development in the science history;
- The Rice statistical distribution's basic features;
- Interconnection between distributions of Rice, Gauss and Rayleigh;
- Solving of two-parameter task by the maximum likelihood method and variants of method of moments .

# The Rice distribution properties



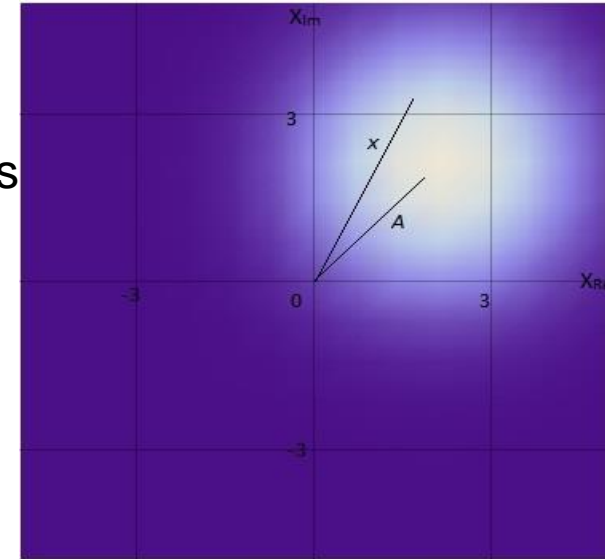
$$\vec{X}(x, \varphi) = \vec{A} + \vec{r} \quad \vec{r} = \vec{r}(r_{\text{Re}}, r_{\text{Im}}) \text{ - Gaussian noise}$$

$$x = \sqrt{x_{\text{Re}}^2 + x_{\text{Im}}^2}$$

$$\overline{r_{\text{Re}}^2} = \overline{r_{\text{Im}}^2} = \sigma^2 \quad \overline{r_{\text{Re}}} = \overline{r_{\text{Im}}} = 0$$

$$A = v$$

$v, \sigma^2$  – Rician parameters



$$P(x|v, \sigma^2) = \frac{x}{\sigma^2} \cdot \exp\left(-\frac{x^2 + v^2}{2\sigma^2}\right) \cdot I_0\left(\frac{xv}{\sigma^2}\right)$$

$$\bar{x} = \sigma \cdot \sqrt{\pi/2} \cdot L_{1/2}\left(-v^2 / 2\sigma^2\right)$$

$$\sigma_x^2 = \overline{x^2} - \bar{x}^2 = 2\sigma^2 + v^2 - \frac{\pi\sigma^2}{2} \cdot L_{1/2}^2\left(-v^2 / 2\sigma^2\right) \quad L_{1/2}(z) = e^{z/2} \left[ (1-z)I_0\left(\frac{-z}{2}\right) - zI_1\left(\frac{-z}{2}\right) \right]$$

# FORMULATION OF THE PROBLEM OF RANDOM RICIAN VALUE TWO-PARAMETER ANALYSIS

Initially determined complex value with amplitude  $\nu$  is distorted by a Gaussian noise,



Amplitude  $x = \sqrt{x_{\text{Re}}^2 + x_{\text{Im}}^2}$  of the resulting signal obeys the Rice statistical distribution with parameters  $\nu$  and  $\sigma^2$ .

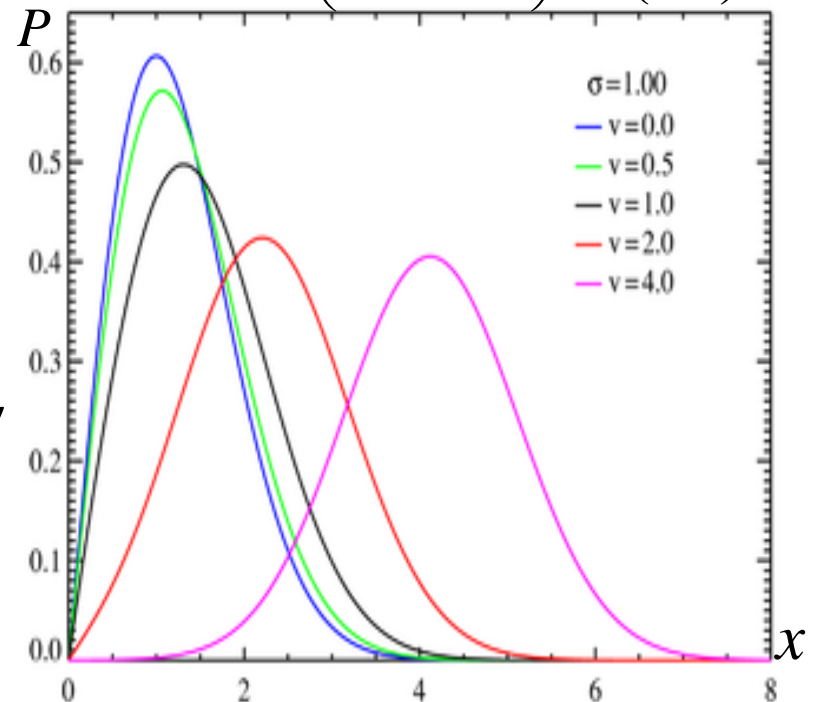
$x_{\text{Re}}, x_{\text{Im}}$  – independent

Gaussian values with dispersion  $\sigma^2$

$$P(x|\nu, \sigma^2) = \frac{x}{\sigma^2} \cdot \exp\left(-\frac{x^2 + \nu^2}{2\sigma^2}\right) \cdot I_0\left(\frac{x\nu}{\sigma^2}\right)$$

**The task** consists in reconstruction of useful signal  $\nu$  against the noise background by joint computing the both Rician parameters

on the basis of measurements of the summary signal  $x_i$  ( $i = 1, \dots, n$ ), without any a priori assumptions concerning value  $\sigma^2$ .



# *Principle stages of the development of Rician data analysis theory*

<b>1880</b>	Rayleigh distribution, introduced by <b>Lord J. Rayleigh, Strutt</b> as a distribution of value $Z = \sqrt{X^2 + Y^2}$ where $X$ and $Y$ are Gaussian values with zero mean
<b>1944</b>	Generalization of the Rayleigh distribution onto the case of nonzero mean of values $X$ and $Y$ ( <b>Stephan O. Rice</b> )
<b>1967</b>	Estimations of the Rician parameters' standard deviations on the basis of Cramer-Rao inequality ( <b>T.R. Benedict, T. T. Soong</b> )
<b>1991/ 1998</b>	Elaboration of one-parameter approximation for the Rician signals' analysis by the method of moments ( <b>K.K. Talukdar, W.D. Lawing</b> )/by the maximum likelihood technique ( <b>J. Sijbers, et al</b> )
<b>2008</b>	A first theoretical consideration of the task of joint signal and noise estimation ( <b>K.Carobbi, M. Cati</b> )
<b>since 2013</b>	Development of a strict theory of the two-parameter Rician signal's analysis as a tool for efficient data processing ( <b>T. Yakovleva, N. Kulberg</b> )

# Comparison of the Rice, Gauss and Rayleigh statistical distributions

$\nu, \sigma$  - the signal and noise parameters

$x$  - value to be analyzed

$$SNR = \nu / \sigma$$

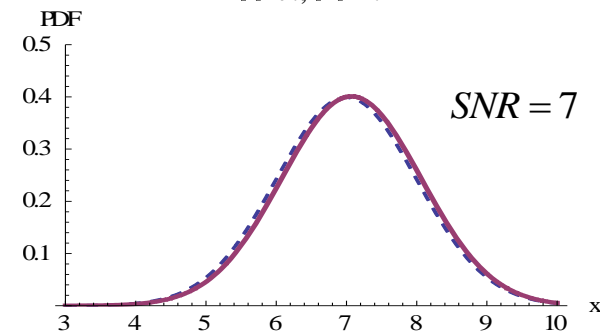
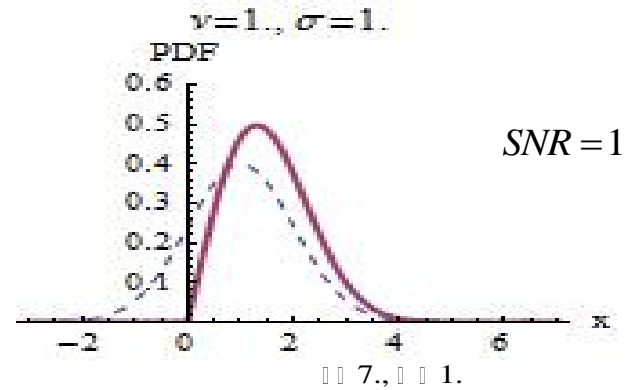
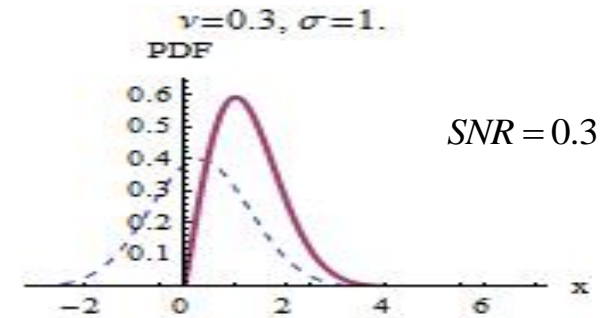
$$P(x|\nu, \sigma^2) = \frac{x}{\sigma^2} \cdot \exp\left(-\frac{x^2 + \nu^2}{2\sigma^2}\right) \cdot I_0\left(\frac{x\nu}{\sigma^2}\right)$$

$$P_G(x|\nu, \sigma^2) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x-\nu)^2}{2\sigma^2}\right)$$

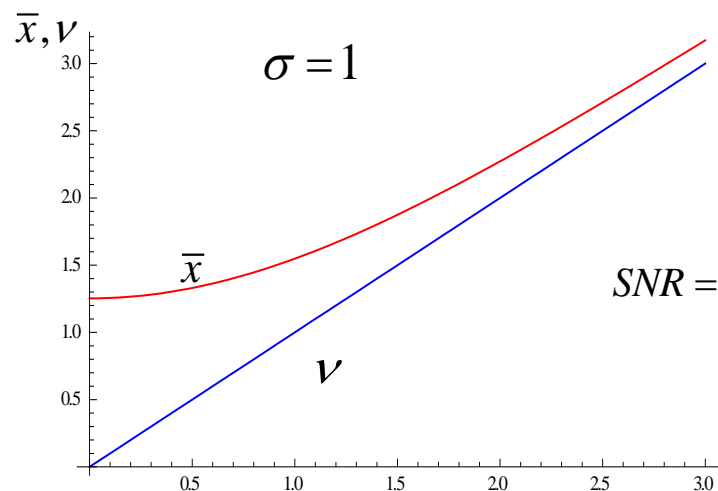
$$I_0(z) \underset{z \gg 1}{\approx} \frac{e^z}{\sqrt{2\pi z}} \left\{ 1 + O\left(\frac{1}{z}\right) \right\} \quad P(x|\nu, \sigma^2) \underset{\frac{\nu}{\sigma} \gg 1}{\rightarrow} P_G(x|\nu, \sigma^2)$$

The difference between the Gauss and Rice distributions is negligible only at sufficiently large values of the signal-to-noise ratio

$$P_R(x|\sigma^2) = \frac{x}{\sigma^2} \cdot \exp\left(-\frac{x^2}{2\sigma^2}\right)$$

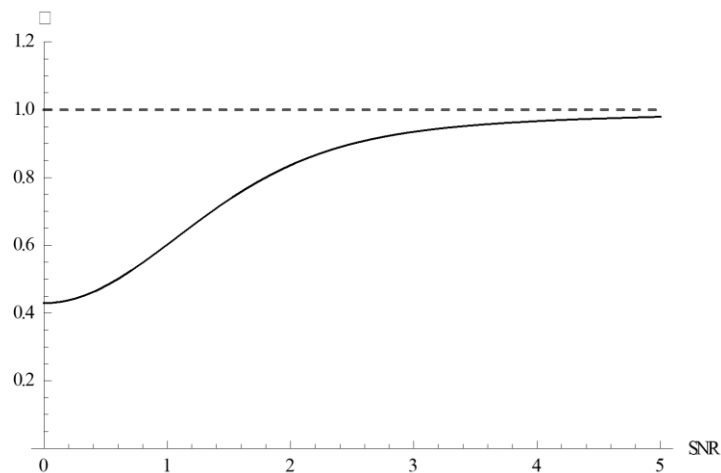


# Nonlinear properties of the Rice distribution

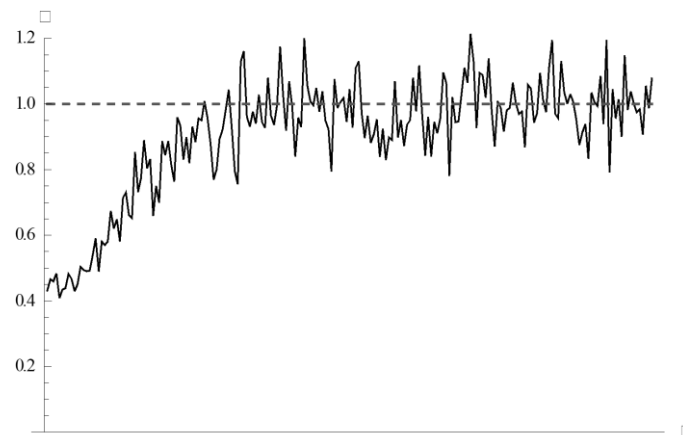


$$\bar{x} = \sigma \cdot \sqrt{\pi/2} \cdot L_{1/2}(-v^2 / 2\sigma^2)$$

$$\sigma_x^2 = \overline{x^2} - \bar{x}^2 = 2 \cdot \sigma^2 + v^2 - \sigma^2 \cdot \frac{\pi}{2} \cdot L_{1/2}^2(-v^2 / 2\sigma^2)$$



**(a) - theory**



**(b) - numerical experiment**



# SOLVING OF THE TWO-PARAMETER TASK BY THE MAXIMUM LIKELIHOOD TECHNIQUE

$V$  - a sought-for amplitude of un-noised signal  $x$  ;

$\sigma^2$  - a dispersion of the Gaussian noise distorting components  $x_{\text{Re}}$  and  $x_{\text{Im}}$

of signal  $x = \sqrt{x_{\text{Re}}^2 + x_{\text{Im}}^2}$

$$P(x|V, \sigma^2) = \frac{x}{\sigma^2} \cdot \exp\left(-\frac{x^2 + V^2}{2\sigma^2}\right) \cdot I_0\left(\frac{xV}{\sigma^2}\right)$$

**The likelihood function** (i.e. a joint probability density function of the events resulting in  $x = x_i$  ( $i = 1, \dots, n$ ), :

$$L\left(\vec{x} | V, \sigma^2\right) = \prod_{i=1}^n P(x_i | V, \sigma) = \prod_{i=1}^n \frac{x_i}{\sigma^2} \cdot \exp\left(-\frac{x_i^2 + V^2}{2\sigma^2}\right) \cdot I_0\left(\frac{x_i V}{\sigma^2}\right)$$

$$\ln L\left(\vec{x} | V, \sigma^2\right) = \sum_{i=1}^n \ln P(x_i | V, \sigma) = \sum_{i=1}^n \left\{ -2 \cdot \ln \sigma - \frac{x_i^2 + V^2}{2 \cdot \sigma^2} + \ln I_0\left(\frac{x_i V}{\sigma^2}\right) \right\}$$

# MAXIMUM LIKELIHOOD (ML) TECHNIQUE FOR SOLVING THE TASK OF TWO-PARAMETER RICIAN DATA ANALYSIS

The key ML equations' system for parameters  $\nu$  and  $\sigma^2$ :

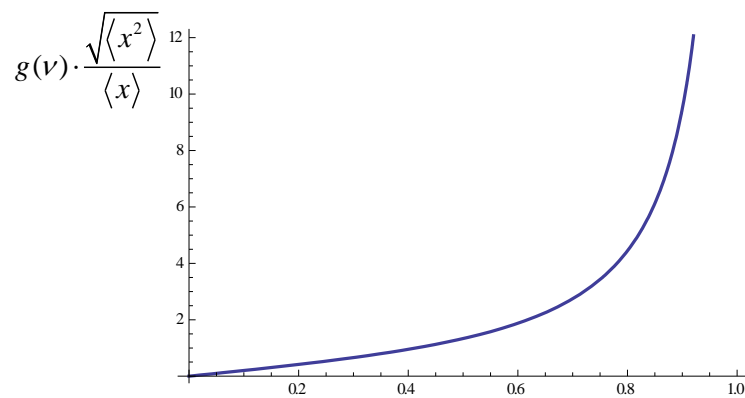
$$\begin{cases} \frac{\partial}{\partial \nu} \ln L(\vec{x} | \nu, \sigma^2) = 0 \\ \frac{\partial}{\partial \sigma} \ln L(\vec{x} | \nu, \sigma^2) = 0 \end{cases} \begin{cases} \nu = \frac{1}{n} \sum_{i=1}^n x_i \cdot \tilde{I}\left(\frac{x_i \nu}{\sigma^2}\right) \\ \sigma^2 = \frac{1}{2 \cdot n} \sum_{i=1}^n (x_i^2 + \nu^2) - \frac{\nu}{n} \sum_{i=1}^n x_i \cdot \tilde{I}\left(\frac{x_i \nu}{\sigma^2}\right) \end{cases} \begin{cases} \nu = \frac{1}{n} \sum_{i=1}^n x_i \cdot \tilde{I}\left(\frac{2 \cdot x_i \cdot \nu}{\langle x^2 \rangle - \nu^2}\right) \\ \sigma^2 = \frac{1}{2} \cdot (\langle x^2 \rangle - \nu^2) \end{cases}$$

**THEOREM: Solution of the ML equation' system exists and is a unique one**

$$g(\nu) = \frac{2 \cdot \langle x \rangle \cdot \nu}{\langle x^2 \rangle - \nu^2}$$

$$\nu = \xi(g(\nu))$$

$$\xi(g) = \frac{1}{n} \sum_{i=1}^n x_i \tilde{I}\left(\frac{x_i}{\langle x \rangle} g\right)$$



$$\tilde{I}(z) = \frac{I_1(z)}{I_0(z)}$$

$$\frac{\nu}{\sqrt{\langle x^2 \rangle}}$$

# Analysis of the second derivative's sign for determination of the character of the likelihood function extremum

$$\frac{\partial^2}{\partial \nu^2} \ln L\left(\vec{x} \mid \nu, \sigma^2\right) = \frac{n}{\sigma^2} \left\{ \frac{\langle x^2 \rangle}{\sigma^2} - 1 - \frac{1}{n \cdot \sigma^2} \sum_{i=1}^n \left[ x_i^2 \cdot \frac{\tilde{I}\left(\frac{x_i \nu}{\sigma^2}\right)}{\left(\frac{x_i \nu}{\sigma^2}\right)} + \tilde{I}^2\left(\frac{x_i \nu}{\sigma^2}\right) \right] \right\}$$

At  $\nu \square 1$ :  $\frac{\partial^2}{\partial \nu^2} \ln L\left(\vec{x} \mid \nu, \sigma^2\right) \square \frac{n}{\sigma^2} \cdot \left(\frac{\langle x^2 \rangle}{2\sigma^2} - 1\right)$

1. Not Rayleigh distribution:  $\langle x^2 \rangle > 2\sigma^2 \implies \frac{\partial^2}{\partial \nu^2} \ln L\left(\vec{x} \mid \nu, \sigma^2\right) > 0$

The trivial solution corresponds to the minimum value of likelihood function

2. Rayleigh distribution:  $\langle x^2 \rangle \leq 2\sigma^2 \implies \frac{\partial^2}{\partial \nu^2} \ln L\left(\vec{x} \mid \nu, \sigma^2\right) \leq 0$

The trivial solution corresponds to the maximum value of likelihood function.

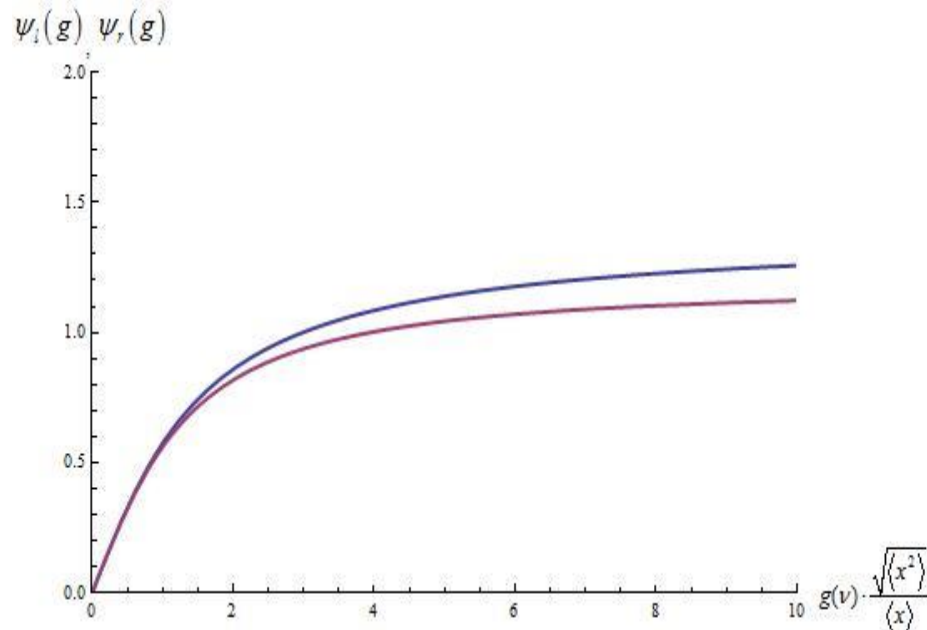
# On proving the theorem concerning the key equation solution

Equation for  $V$  is transformed to equation

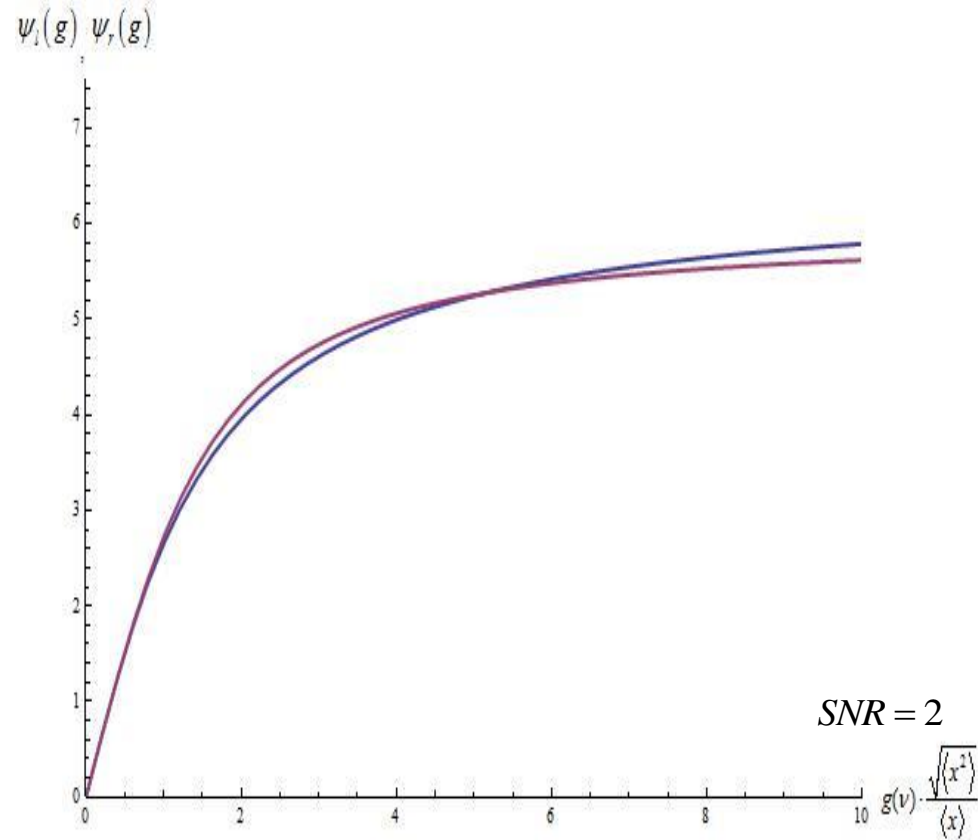
$$\xi(g) = \frac{1}{n} \sum_{i=1}^n x_i \tilde{I} \left( \frac{x_i}{\langle x \rangle} g \right) \quad (*)$$

for variable  $g(v) = \frac{2 \cdot \langle x \rangle \cdot v}{\langle x^2 \rangle - v^2}$ :

$$\psi_l(g) = \xi(g) \quad \psi_r(g) = \frac{1}{n} \sum_{i=1}^n x_i \tilde{I} \left( \frac{x_i}{\langle x \rangle} g \right)$$



The particular case of Rayleigh distribution

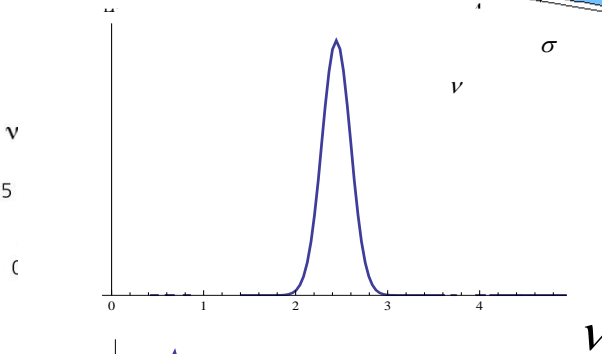
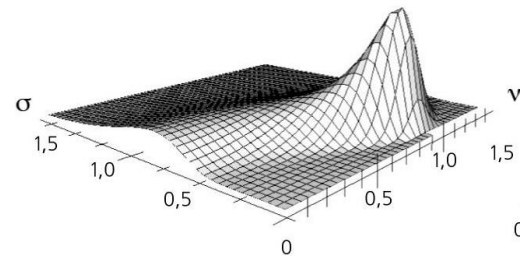
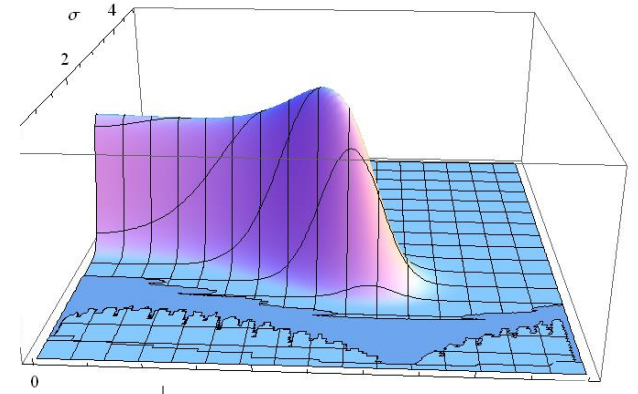
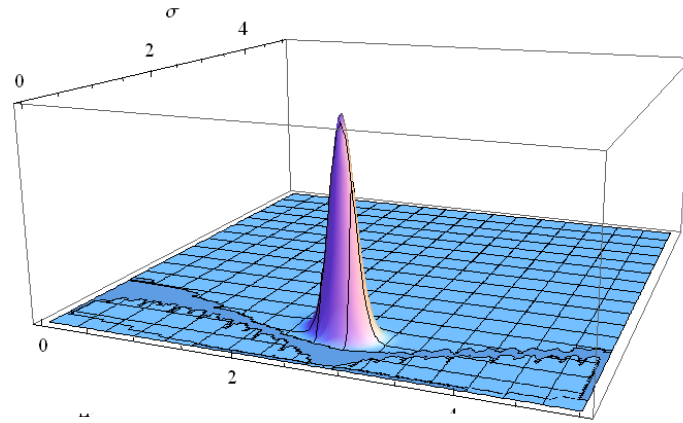
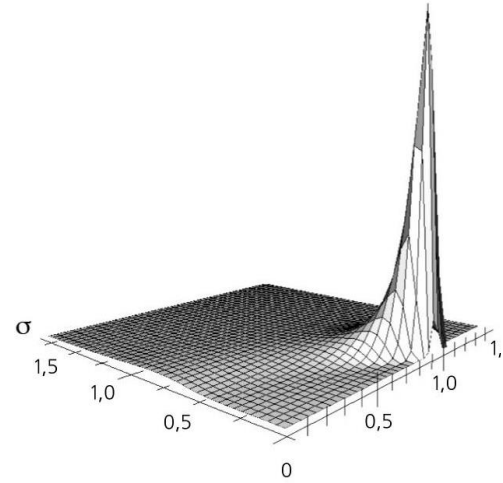


The common case of Rice distribution

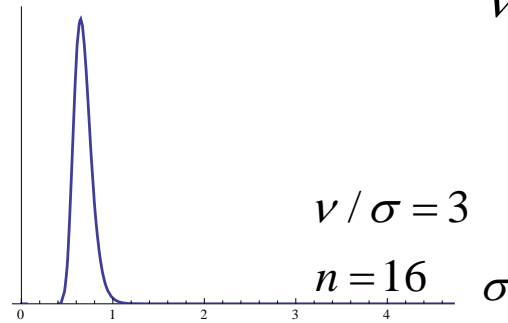
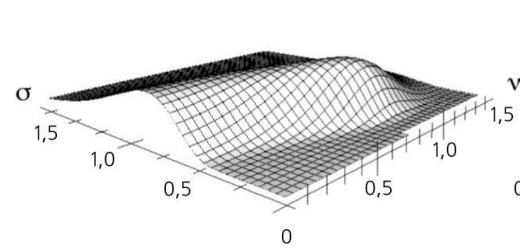
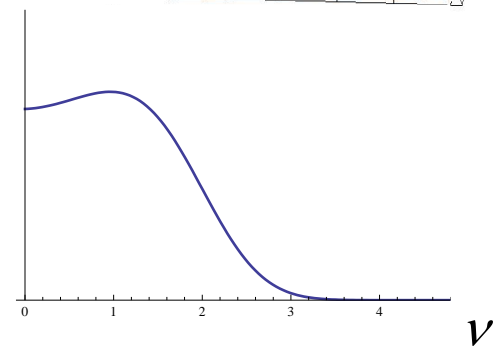
$$v = 0$$

$$\sigma^2 = \frac{1}{2} \cdot \langle x^2 \rangle$$

# RICIAN LIKELIHOOD FUNCTION AS DEPENDENT UPON THE SIGNAL-TO-NOISE RATIO

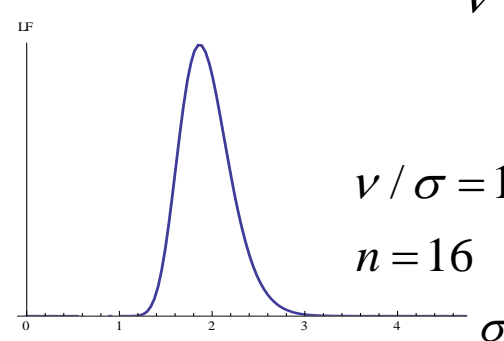


$\sigma = \text{const}$



$\nu/\sigma = 3$   $\nu = \text{const}$

$n = 16$

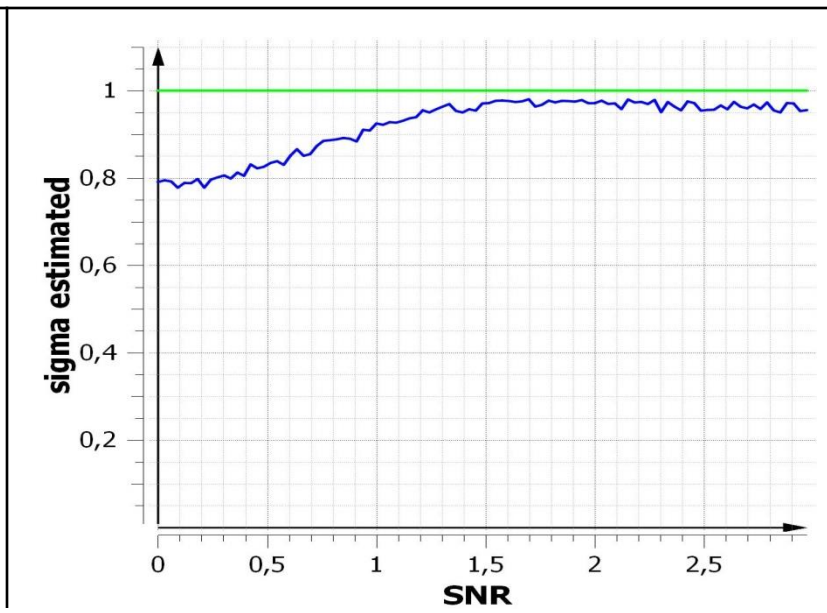
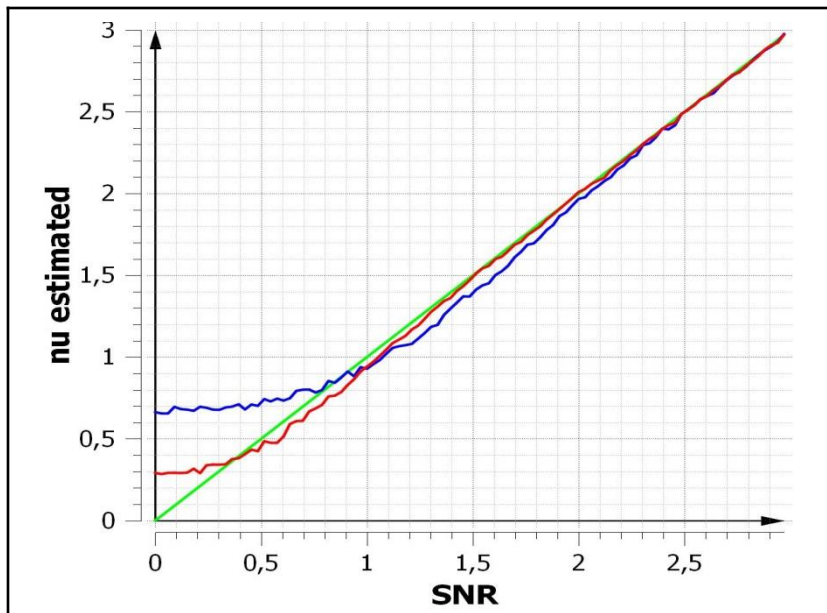


$\nu/\sigma = 1$

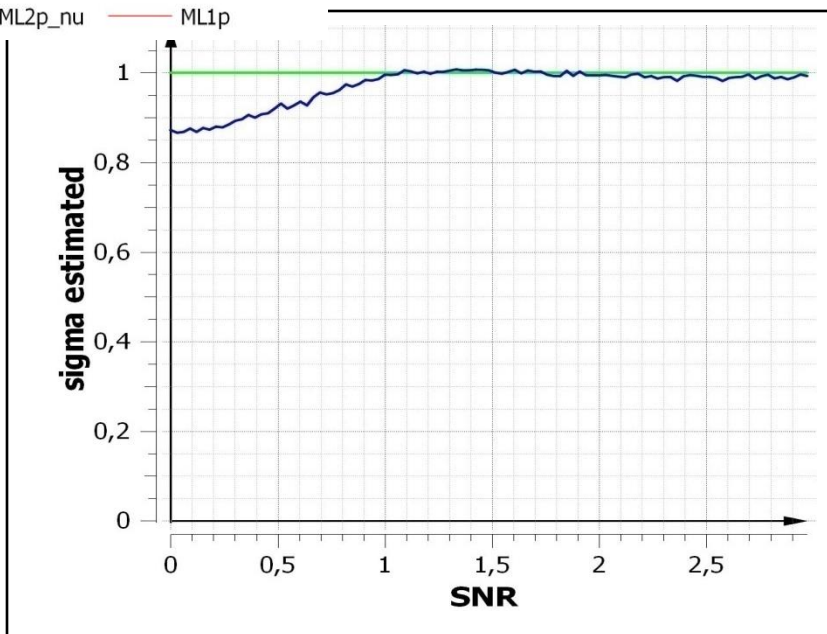
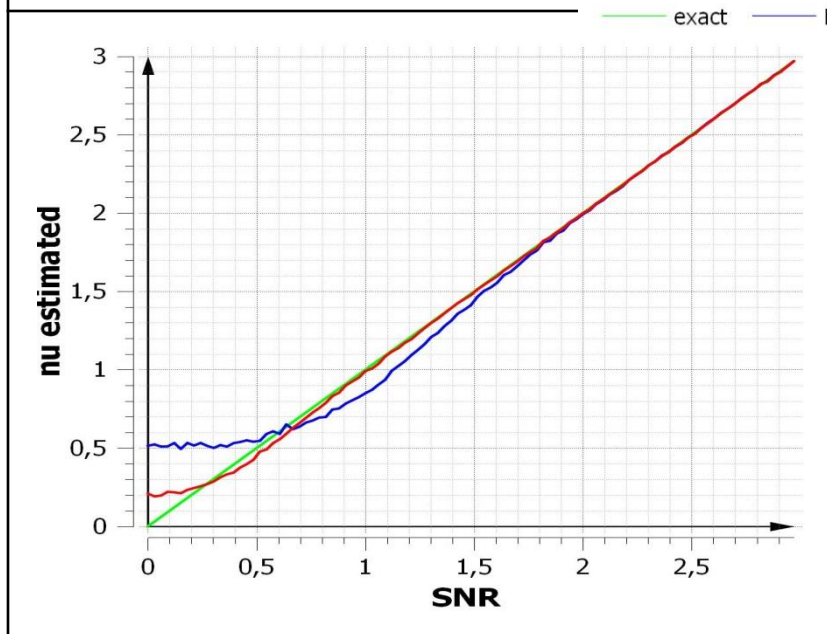
$n = 16$

$\nu/\sigma = 10; 4; 2$   
 $n = 3$

# РЕЗУЛЬТАТЫ ЧИСЛЕННОГО МОДЕЛИРОВАНИЯ РЕШЕНИЯ ЗАДАЧИ МЕТОДОМ МАКСИМУМА ПРАВДОПОДОБИЯ



$n=16$



$n=64$

$\sigma = 1$

$N_{aver} = 1000$

# Variants of the Method of Moments at solving the Task of Two-parameter Analysis of Rician Data

1. The method of lower even-numbered moments (MM24);
2. The method of lower moments (MM12).

## Two-parameter method MM24

$$\begin{cases} \overline{x^2} = 2 \cdot \sigma^2 + \nu^2 \\ \overline{x^4} = 8 \cdot \sigma^4 + 8 \cdot \sigma^2 \cdot \nu^2 + \nu^4 \end{cases} \implies \begin{cases} \nu^2 = \langle x^2 \rangle \sqrt{1-t} \\ \sigma^2 = \frac{\langle x^2 \rangle}{2} (1 - \sqrt{1-t}) \end{cases}$$
$$t = \frac{\langle x^4 \rangle}{(\langle x^2 \rangle)^2} - 1 \quad 0 < t \leq 1$$

At  $\nu = 0 \quad t = 1 \quad \sigma^2 = \frac{\langle x^2 \rangle}{2}$

## Two-parameter method MM12

1-st moment of rician value:

$$\bar{x} = \sigma \cdot \sqrt{\pi / 2} \cdot L_{1/2} \left( -v^2 / 2\sigma^2 \right)$$

$L_{1/2}$  - Laguerre polynomial:

$$L_{1/2}(z) = e^{z/2} \left[ (1-z) I_0 \left( \frac{-z}{2} \right) - z I_1 \left( \frac{-z}{2} \right) \right]$$

$$\begin{cases} \sigma \cdot \sqrt{\pi / 2} \cdot e^{-\frac{v^2}{4\sigma^2}} \left[ \left( 1 + \frac{v^2}{2\sigma^2} \right) I_0 \left( \frac{v^2}{4\sigma^2} \right) + \frac{v^2}{2\sigma^2} I_1 \left( \frac{v^2}{4\sigma^2} \right) \right] = \bar{x} \\ 2\sigma^2 + v^2 = \bar{x}^2 \end{cases}$$

$$\begin{cases} \sqrt{\frac{\pi}{2}} \cdot \sigma^2 \cdot e^{-\frac{r}{2}} \cdot \left[ (1+r) I_0 \left( \frac{r}{2} \right) + r I_1 \left( \frac{r}{2} \right) \right] = \langle x \rangle \\ 2\sigma^2 (1+r) = \langle x^2 \rangle \end{cases}$$

$$r = \frac{v^2}{2\sigma^2}$$

**MM12 – method: equation for  $r$**

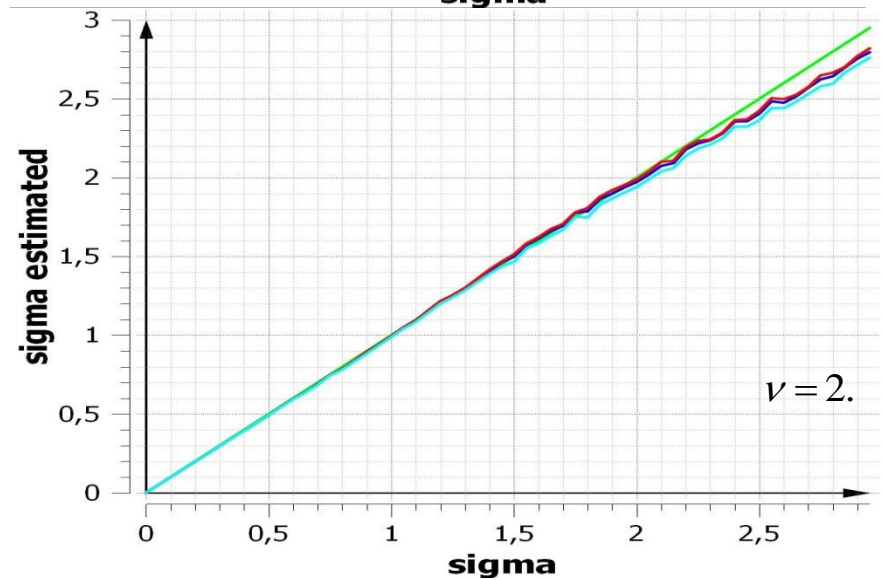
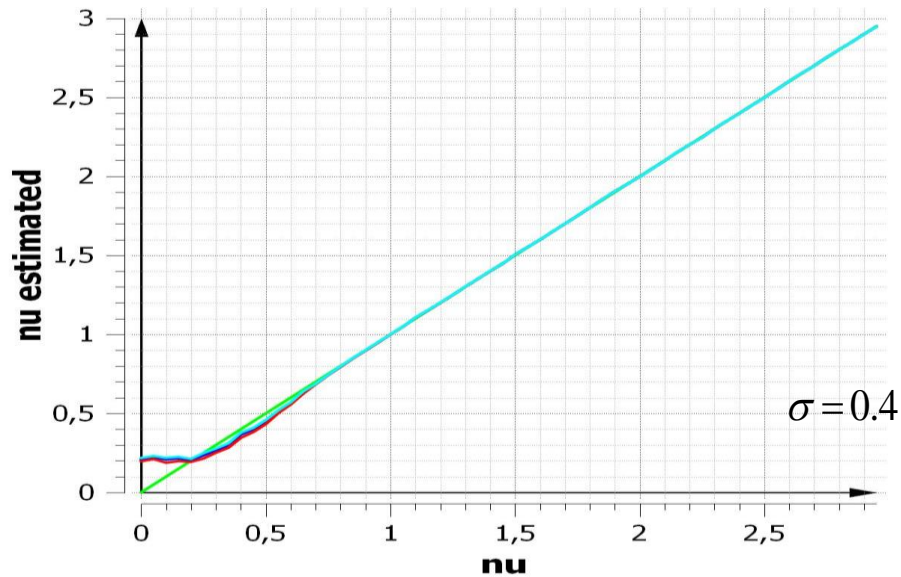
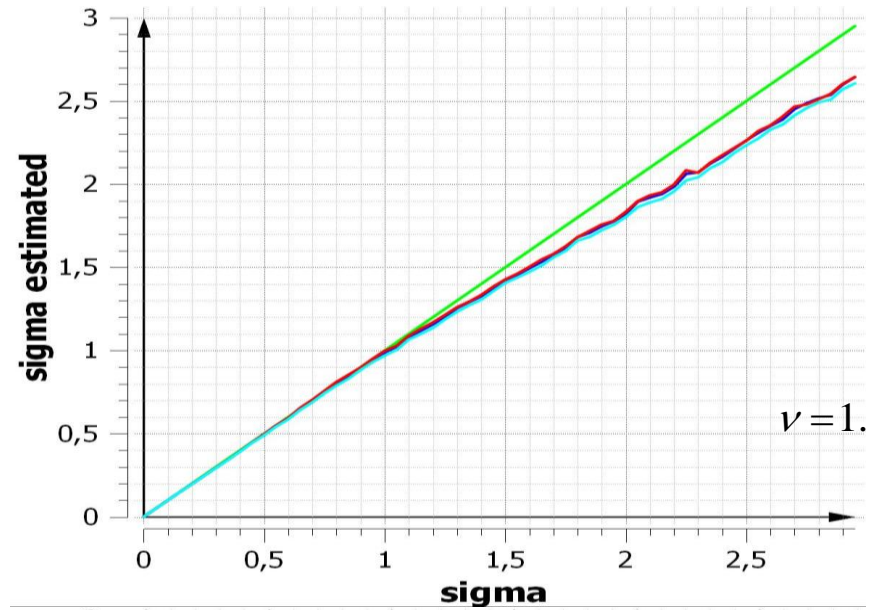
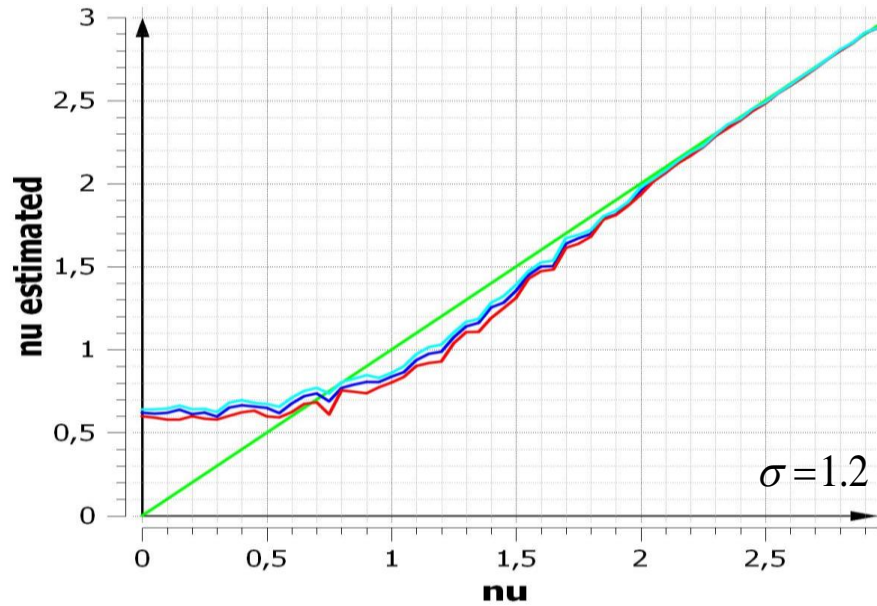
$$\sqrt{\frac{\pi}{4}} \cdot \langle x^2 \rangle \cdot \sqrt{1+r} \cdot e^{-\frac{r}{2}} \cdot I_0 \left( \frac{r}{2} \right) \left[ 1 + \frac{r}{(1+r)} \cdot \tilde{I} \left( \frac{r}{2} \right) \right] = \langle x \rangle$$

$$v = \sqrt{\frac{r}{1+r}} \sqrt{\langle x^2 \rangle}$$

$$\sigma^2 = \frac{\langle x^2 \rangle}{2(1+r)}$$



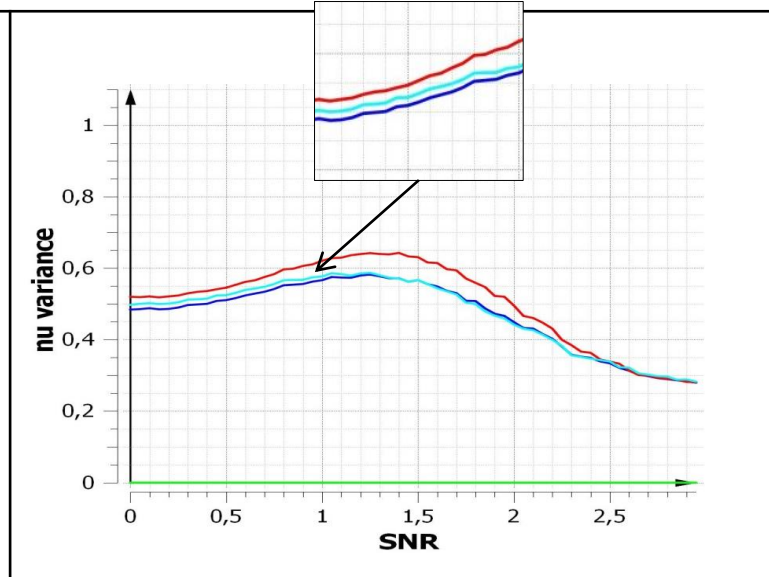
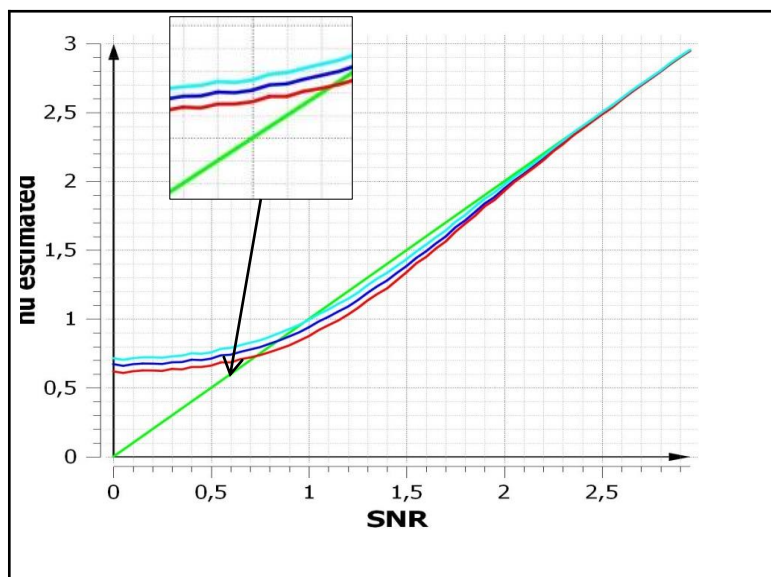
# Results of the task digital solution by the elaborated techniques



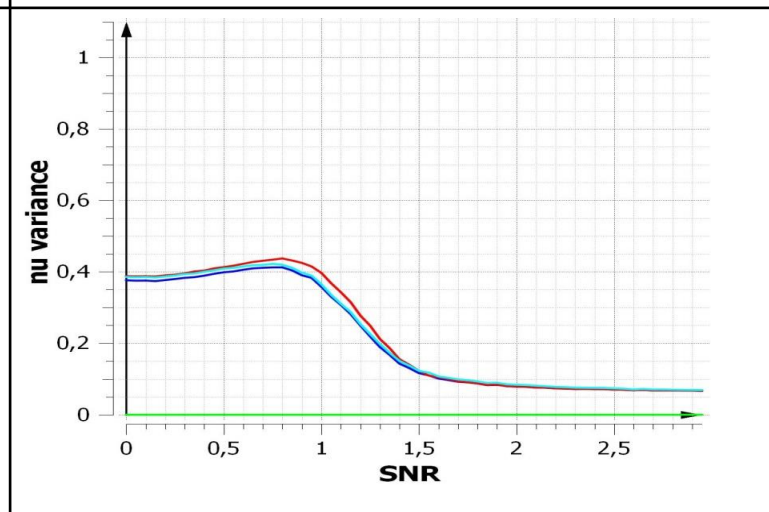
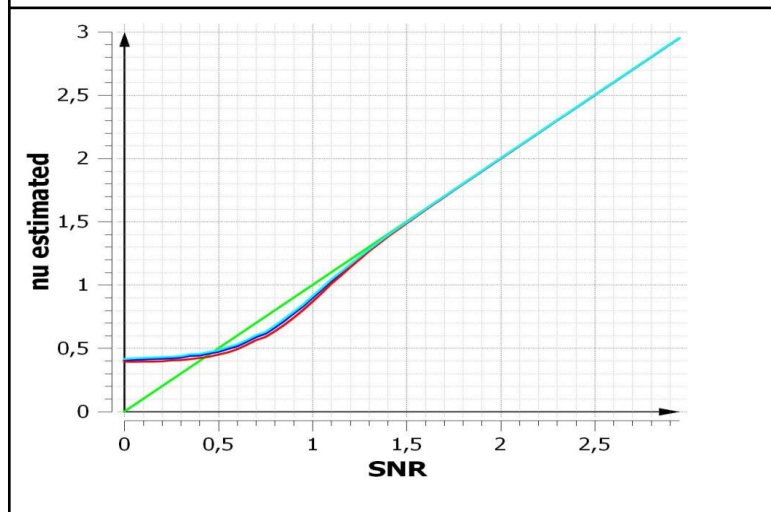
— exact — ML2p\_nu — MM12 — MM24

$n = 64$   $N_{av} = 500$

# The plots of parameter $\nu$ (left) and its standard deviation (right), calculated by methods ML, MM12, MM24



$n=16$

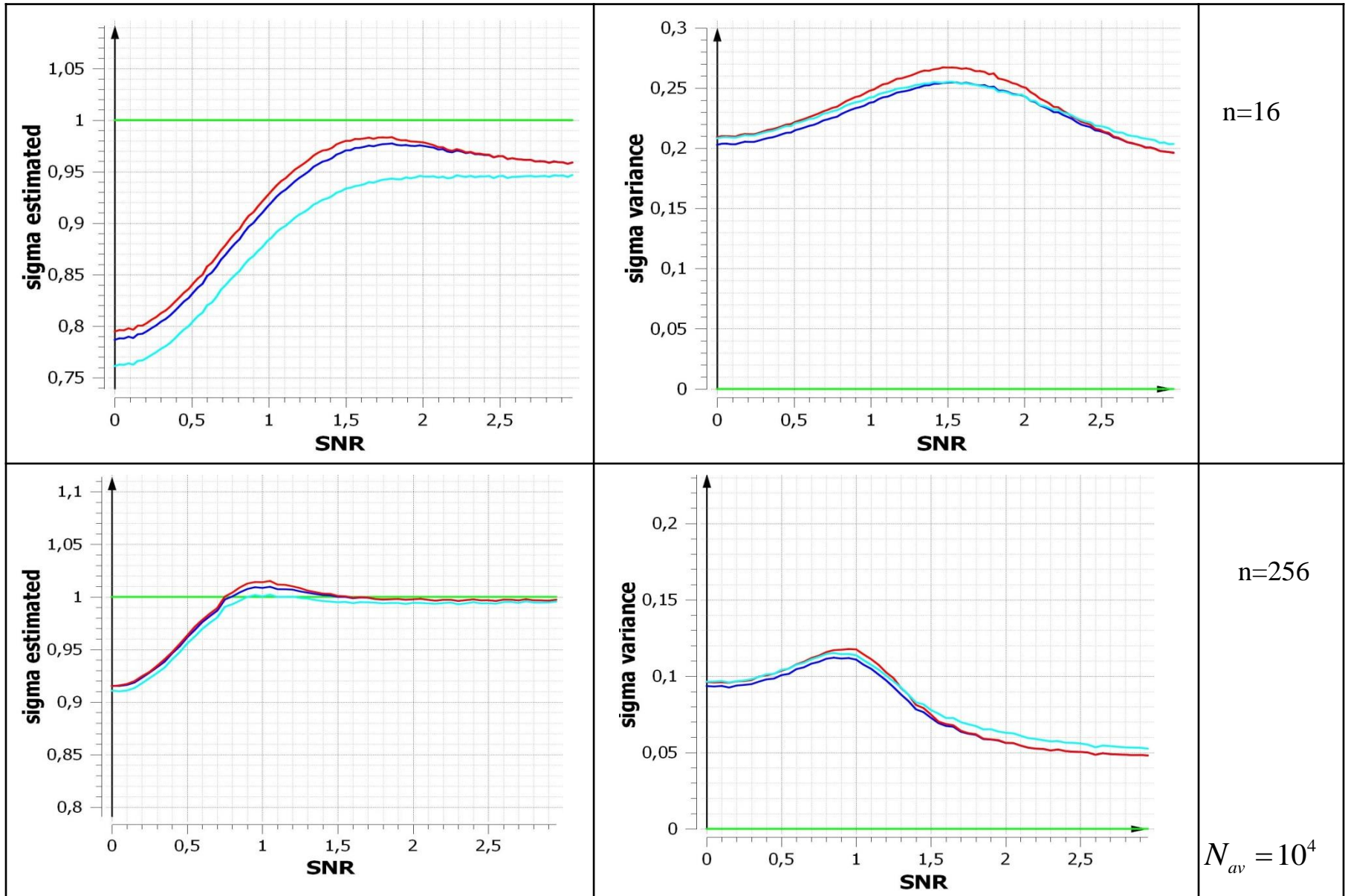


$n=256$

$\sigma = 1 \quad N_{av} = 10^4$

— exact — ML2p\_nu — MM12 — MM24

# The plots for calculated parameter $\sigma$ (on the left) and their mean square deviations (on the right)



— exact — ML2p\_nu — MM12 — MM24

## TWO-PARAMETER COMBINED ML-MM METHOD

$\nu, \sigma^2$  – Rician parameters

$$\begin{cases} \nu = \frac{1}{n} \sum_{i=1}^n x_i \cdot \tilde{I} \left( \frac{x_i \cdot \nu}{\sigma^2} \right) \\ \frac{1}{n} \sum_{i=1}^n x_i = \sigma \sqrt{\frac{\pi}{2}} \cdot L_{1/2} \left( -\frac{\nu^2}{2\sigma^2} \right) \end{cases} \Rightarrow 2\sqrt{r/\pi} \cdot \frac{\langle x \rangle}{L_{1/2}(-r)} = \frac{1}{n} \sum_{i=1}^n x_i \tilde{I} \left( \frac{x_i \sqrt{\pi \cdot r}}{\langle x \rangle} L_{1/2}(-r) \right)$$

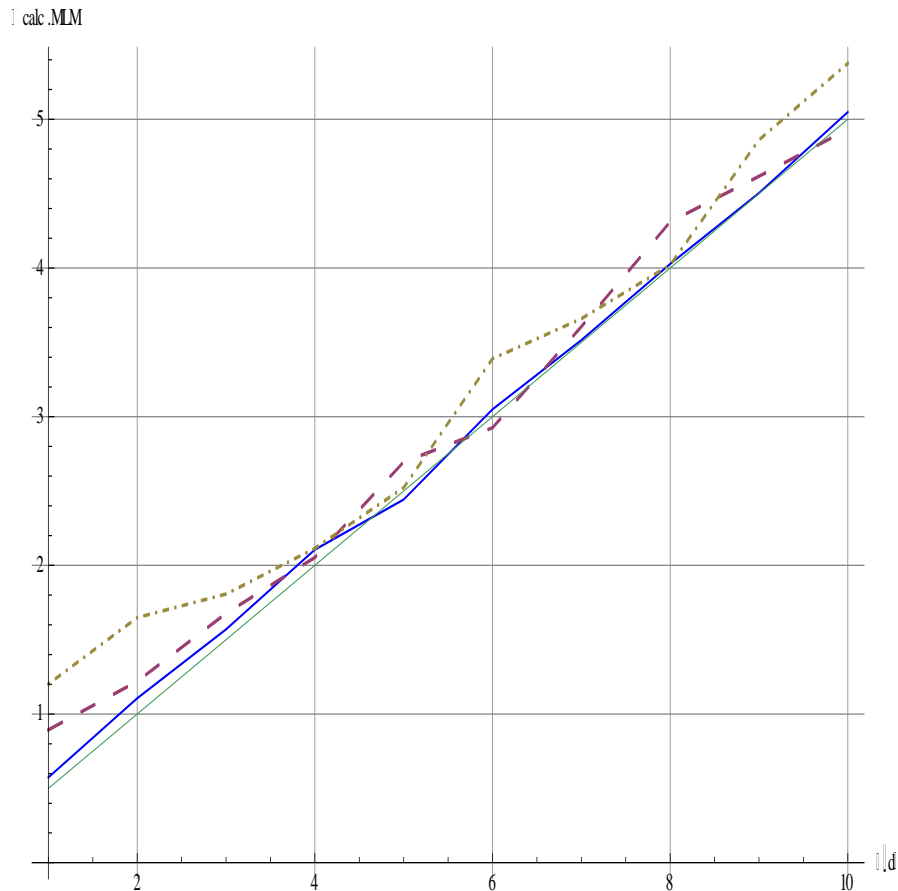
$L_q(z)$  - the Laguerre polynomial  $r = \frac{\nu^2}{2\sigma^2}$   $\nu = \sqrt{2\sigma^2 r}$

## TWO-PARAMETER METHOD OF MOMENTS MM13

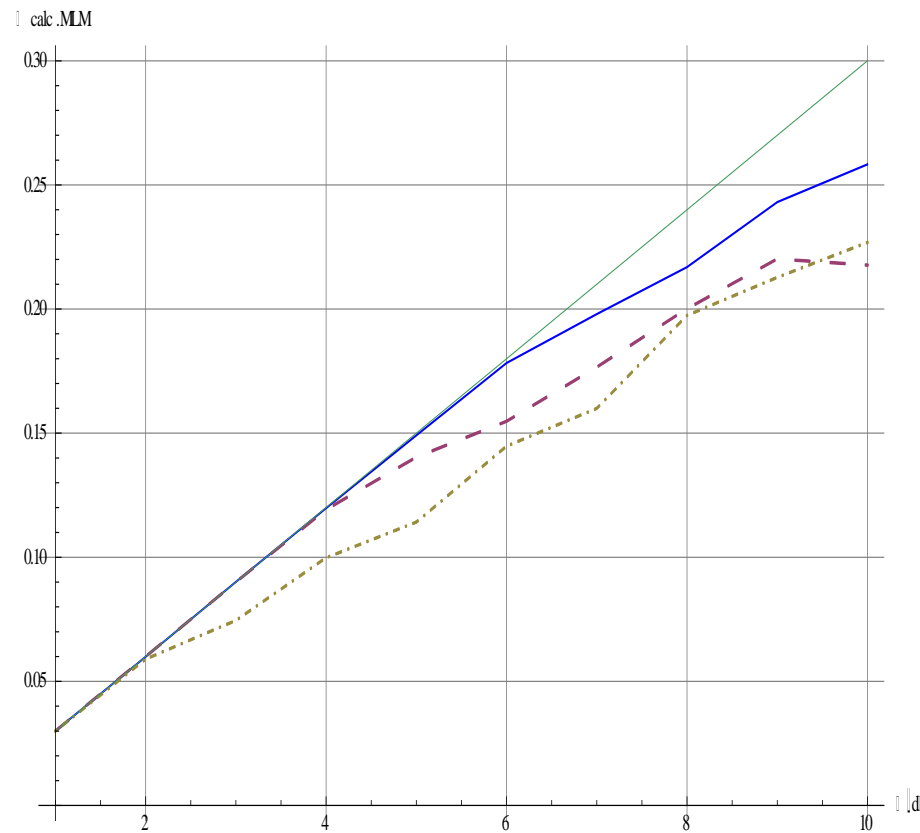
$$\begin{aligned} \bar{x} &= \sigma \cdot \sqrt{\frac{\pi}{2}} \cdot L_{1/2} \left( -\nu^2/2\sigma^2 \right), \\ \bar{x^3} &= 3 \cdot \sigma^3 \cdot \sqrt{\frac{\pi}{2}} \cdot L_{3/2} \left( -\nu^2/2\sigma^2 \right). \end{aligned} \Rightarrow \langle x \rangle^3 \cdot {}_1F_1 \left( -\frac{3}{2}; 1; -r \right) = \langle x^3 \rangle \cdot \frac{\pi}{6} \cdot {}_1F_1 \left( -\frac{1}{2}; 1; -r \right),$$

$$\sigma = \langle x \rangle \sqrt{\frac{2}{\pi}} / {}_1F_1 \left( -\frac{1}{2}; 1; -r \right) \quad {}_1F_1 \text{ -confluent hypergeometric function of the 1-st order, or Kummer's function}$$

# Computer simulation results for ML-MM method



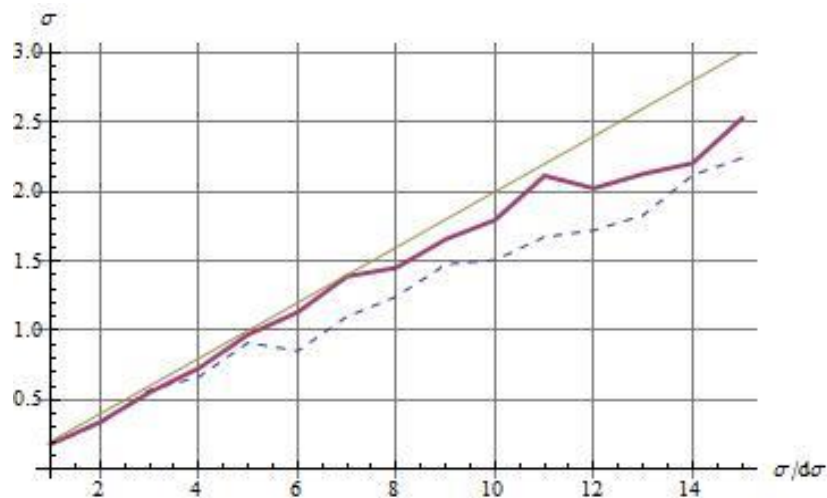
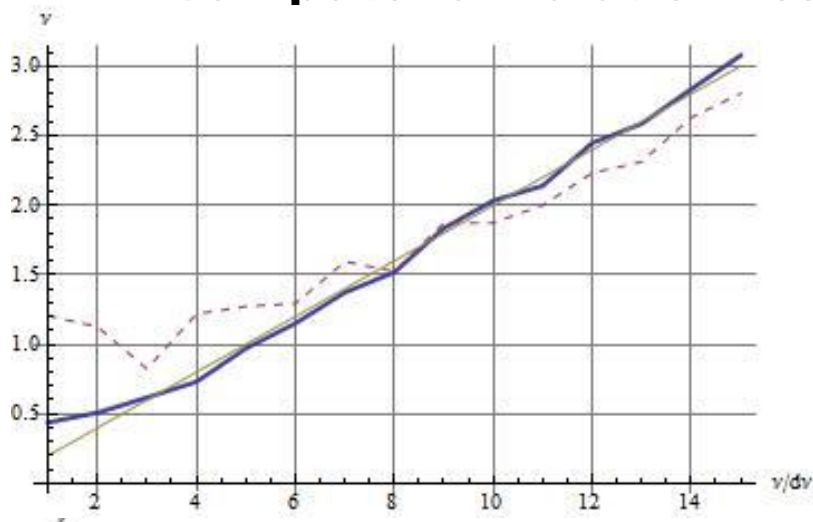
$\nu = 0,5$  (solid line); 1 (dashed line); 1,5 (dotted-dashed line)



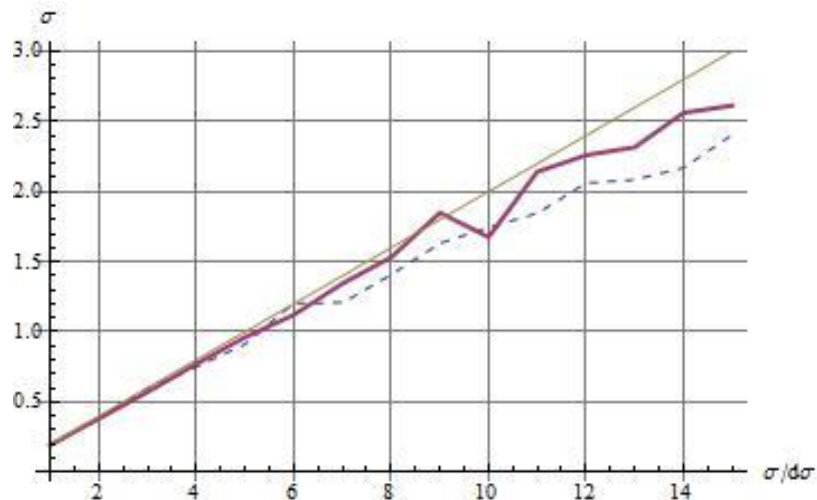
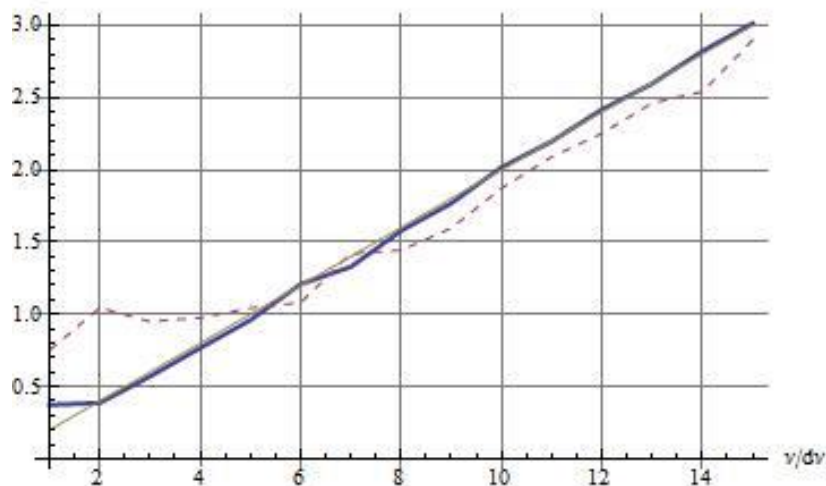
$\nu = 3.0$  (solid line); 2.0 (dashed line); 1.0 (dotted-dashed line)

n=4

# Computer simulation results for MM13 method



n=8



n=16

$G = 0,6$  (solid line); 1,5 (dashed line)

$\nu = 3.0$  (solid line); 1,5 (dashed line).

# The meaningful peculiarities of the proposed methods of two-parameter analysis

- The possibility of computing the both sought-for parameters of the task by computer algebra techniques without any additional calculative capacities if compared with the traditional one-parameter approximation.
- The applicability of the elaborated techniques of the rician data analysis for solving the nonlinear tasks in the absence of any restrictions connected with the *a-priori* suppositions on the parameters of the task.
- The elaborated techniques may be applied in data processing systems with the priority of operation in a real time mode.

# Conclusion

- Concept and methods of the ***two-parameter analysis based upon the both Rician parameters joint computing*** have been developed **to optimize the computer algebra techniques for solving the task** of separation of the informative and the noise signal's components, namely:

solving ***the system of two nonlinear equations with two unknown variables can be reduced to solving just one equation for only one unknown variable;***

- The proposed approach provides a new efficient tool for ***accurate reconstruction of the Rician signal's informative component against the noise background.***



*Thank you very  
much for your  
attention*