

# *Model order selection criteria: comparative study and applications*

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# Outline

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- Model order selection.
- Correlation matrix.
- Akaike Information Criterion – AIC. Minimum Description Length – MDL, Minka Bayesian Information Criterion – MIBS.
- Investigations – basic properties.
- Time-frequency parametric spectrum.
- Switching of the condenser banks – parameter estimation of the current.

# Model order – why?

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- Find the appropriate dimensionality which fits a given set of observations
  - Examples:
    - choice of degree for polynomial regression, order of multi-step Markov chain
    - estimating the number of signals in multi-channel time series
    - AR model order selection
- estimation of the signal and subspace dimension → correct estimates of the signal parameters

# Correlation matrix

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- Signal samples

$$\mathbf{x} = \begin{bmatrix} x(0) \\ x(1) \\ \vdots \\ x(N-1) \end{bmatrix}$$

- Correlation matrix

$$\mathbf{R}_x = \mathcal{E} \{ \mathbf{x} \cdot \mathbf{x}^{*T} \} = \begin{bmatrix} R_x(0,0) & R_x(0,1) & \cdots & R_x(0,N-1) \\ R_x(1,0) & R_x(1,1) & \cdots & R_x(1,N-1) \\ \vdots & \vdots & \ddots & \vdots \\ R_x(N-1,0) & R_x(N-1,1) & \cdots & R_x(N-1,N-1) \end{bmatrix}$$

- Eigenvalues

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_p$$

# AIC & MDL

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*second term makes AIC  
inconsistent estimator*

$$\text{AIC} = -2 \log f(X|\hat{\Theta}) + 2k$$

$$\text{MDL} = -\log f(X|\hat{\Theta}) + \frac{1}{2}k \log N$$

# Derivation of AIC & MDL

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$$\text{AIC}(k) = -2 \log \left( \frac{\prod_{i=k+1}^p \lambda_i^{\frac{1}{p-k}}}{\frac{1}{p-k} \sum_{i=k+1}^p \lambda_i} \right)^{(p-k)N} + 2k(2p - k)$$

$$\text{MDL}(k) = -\log \left( \frac{\prod_{i=k+1}^p \lambda_i^{\frac{1}{p-k}}}{\frac{1}{p-k} \sum_{i=k+1}^p \lambda_i} \right)^{(p-k)N} + \frac{1}{2}k(2p - k) \log N$$

## Criterion of Minka (MIBS)

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- Based on Probabilistic PCA Model
- Operates on eigenvalues of the data correlation matrix
- Fast & accurate
- Scores models according to probability they assign the observed data
- Similar to **Bayesian Classification**
- Preference for simpler models

# Derivation of MIBS

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$$p(\mathbf{X}|q) = p(U) \left( \prod_{j=1}^q \lambda_j \right)^{-N/2} \hat{\sigma}_{\text{ML}}^{-N(p-q)} \cdot (2\pi)^{(m+q)/2} |A_z|^{-1/2} N^{-q/2}$$

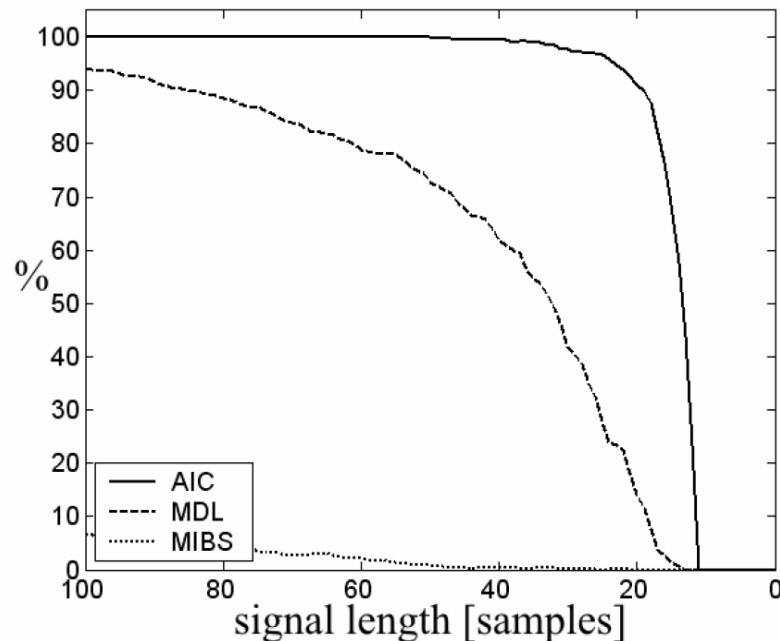
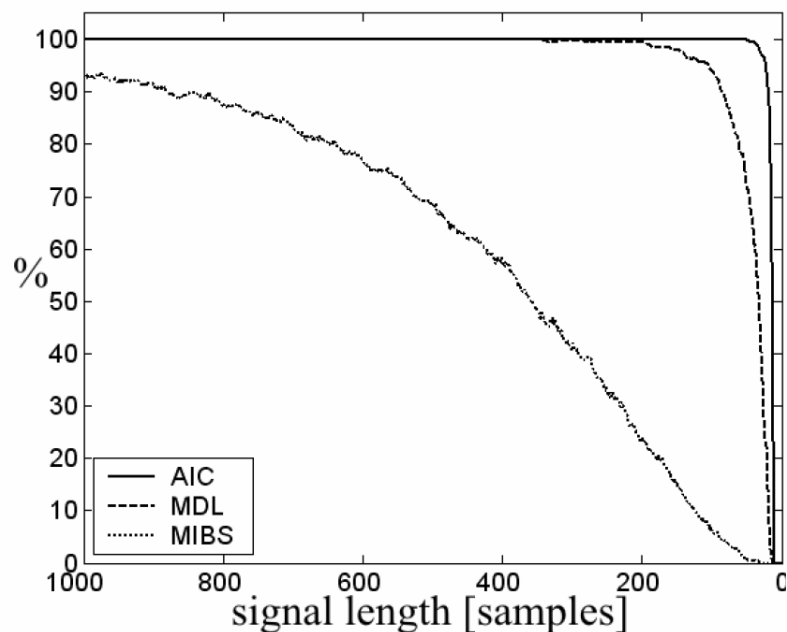
$$p(U) = 2^{-q} \prod_{j=1}^q \Gamma((p-j+1)/2) \pi^{\frac{-(p-i+1)}{2}}$$

$$|A_z| = \prod_{i=1}^q \prod_{j=i+1}^p N(\hat{\lambda}_j^{-1} - \hat{\lambda}_i^{-1})(\lambda_i - \lambda_j)$$

- Laplace method for approximation of integrals in Bayesian statistics

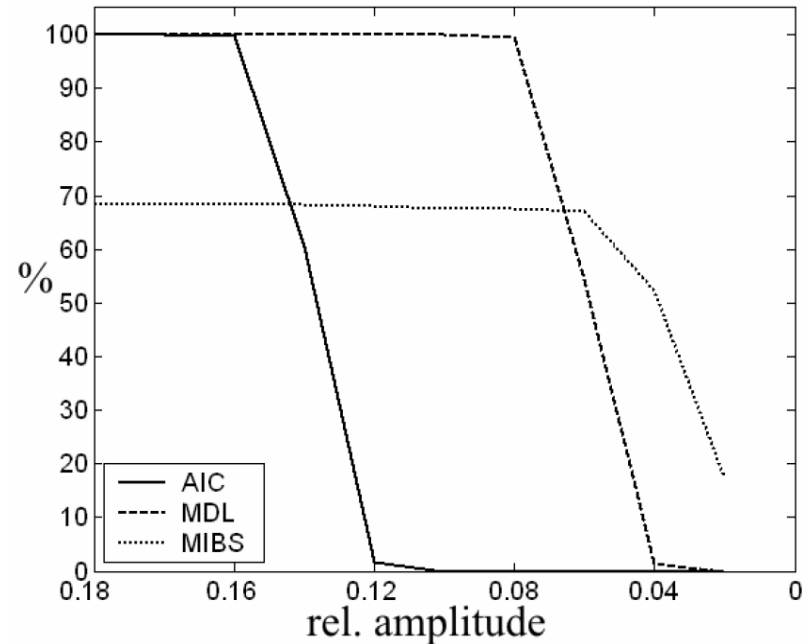
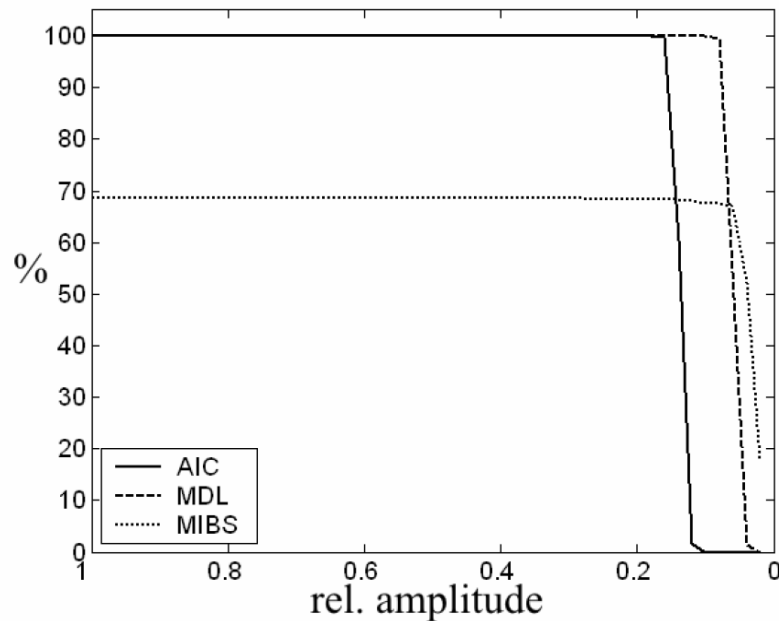


# INVESTIGATIONS



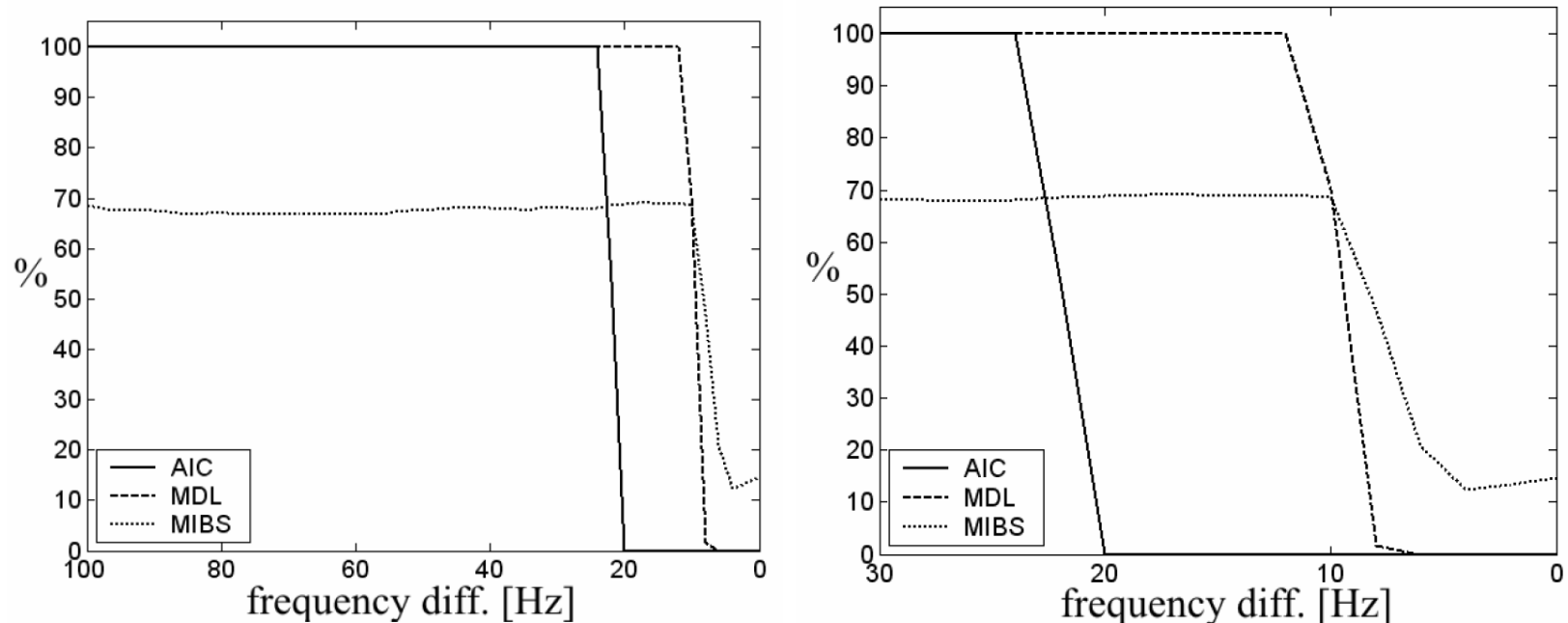
- Accuracy  $\leftrightarrow$  window length
  - 2 sinusoids 50Hz & 150Hz, SNR=20dB, 1000 real.

# INVESTIGATIONS



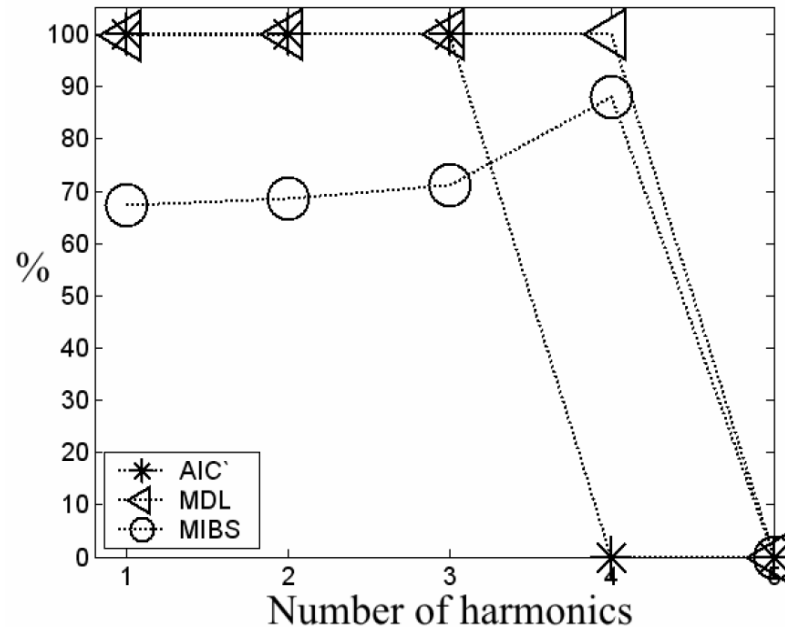
- Accuracy ↔ masking
  - Basic component 50Hz with amplitude one & 150Hz decreasing amplitude, SNR=20dB

# INVESTIGATIONS



- Accuracy  $\leftrightarrow$  difference in frequency
  - Basic component 50Hz with amplitude one & second with frequency from 80 to 50 Hz, SNR=20dB

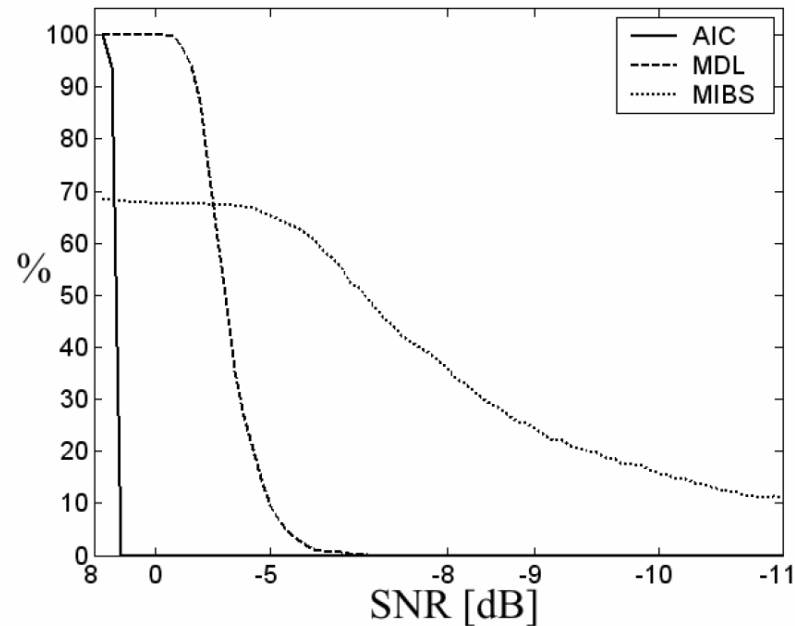
# INVESTIGATIONS



- Accuracy ↔ number of components
  - 50, 100, 150, 200, 250 Hz, SNR=20dB

# INVESTIGATIONS

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- Accuracy  $\leftrightarrow$  Gaussian noise
  - Components 50, 150 Hz, SNR from 8 to -11 dB

# Time-frequency parametric spectrum - MUSIC

- Eigenfilter  $E_i(z) = e_i[0] + e_i[1]z^{-1} + \dots + e_i[N-1]z^{-(N-1)}$

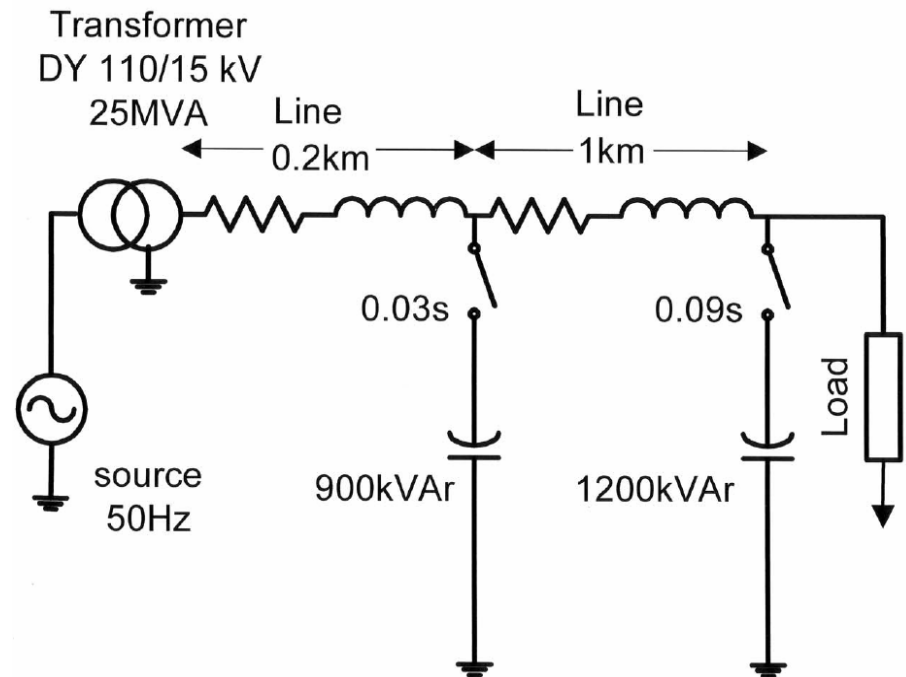
$$\mathbf{w} = \begin{bmatrix} 1 & e^{j\omega_i} & \dots & e^{j(N-1)\omega_i} \end{bmatrix}^T$$

$$\mathbf{w}^{*T} \mathbf{E}_{noise} \mathbf{E}_{noise}^{*T} \mathbf{w} = \sum_{i=M+1}^N \mathbf{w}^{*T} \mathbf{e}_i \mathbf{e}_i^{*T} \mathbf{w}$$

- The polynomial has  $M$  double roots lying on the unit circle, correspond to the signal frequencies.  $\hat{P}(e^{j\omega}) = \left[ \sum_{i=M+1}^N E_i(z) E_i^*(1/z^*) \right]^{-1} \Big|_{z=e^{j\omega}}$

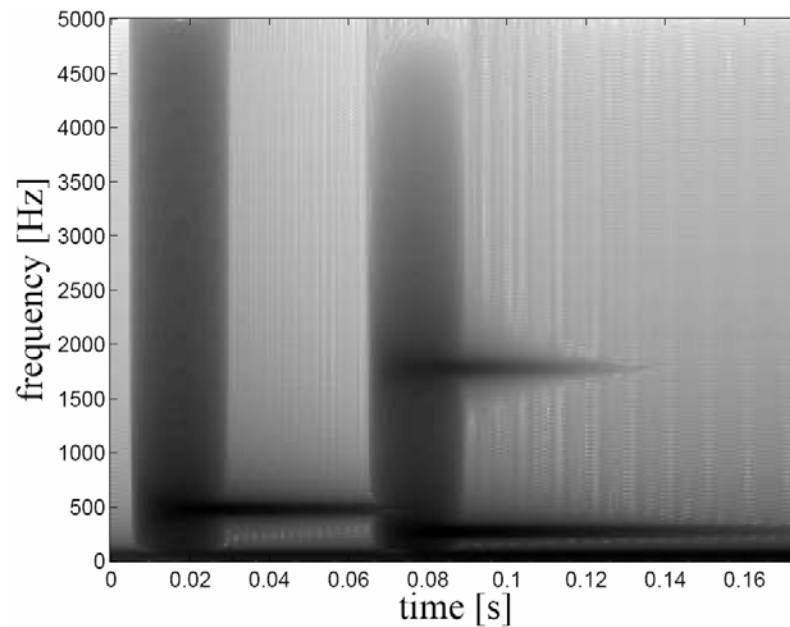
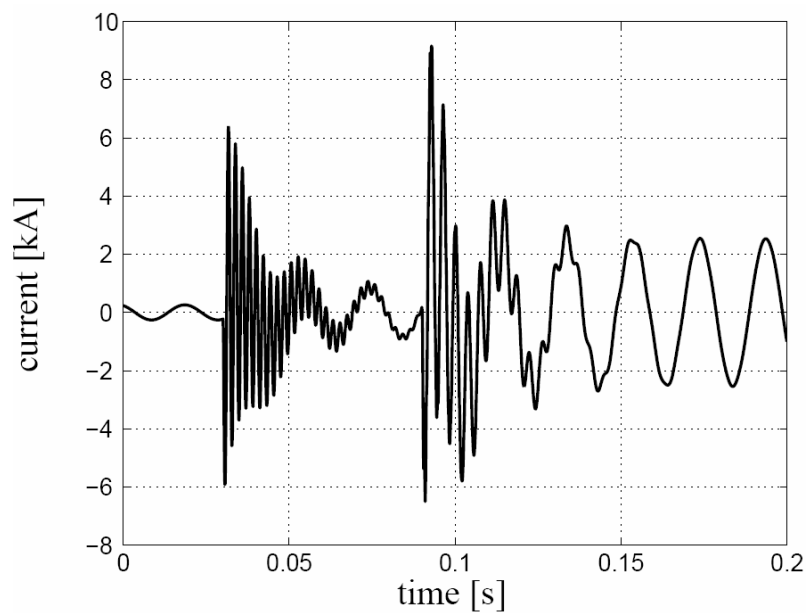
# Switching of the condenser banks

- sampling frequency 10 kHz
- analysis window 100 samples (0.01 s).
- The 1<sup>st</sup> condenser bank was switched on at  $t = 0.03$  s and the 2<sup>nd</sup> at  $t=0.09$  s.



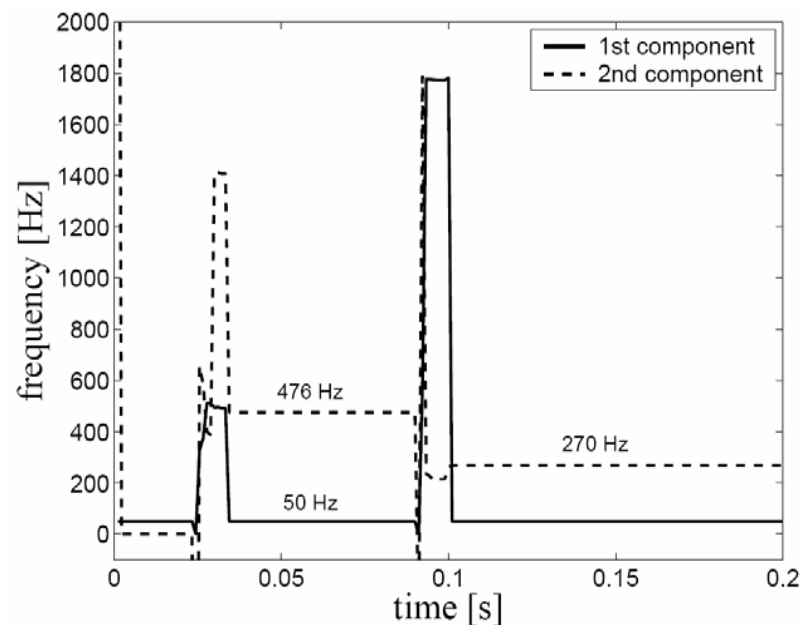
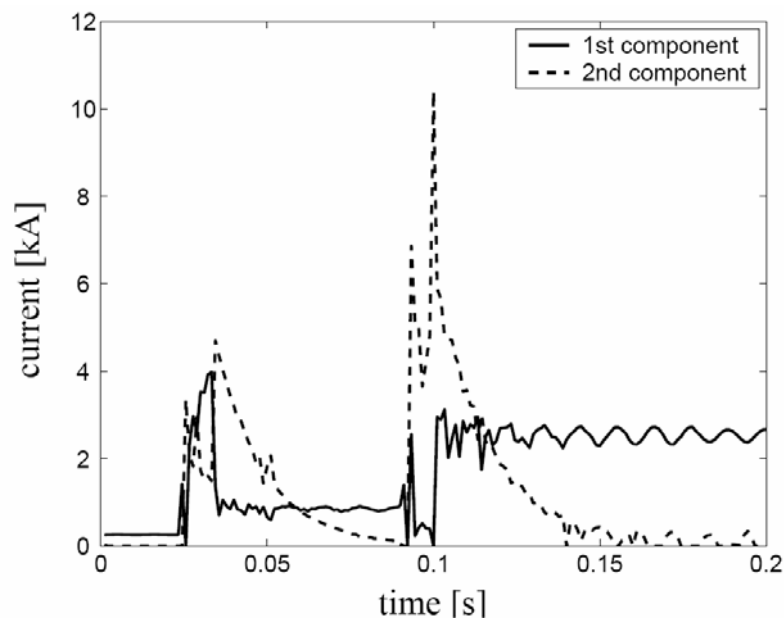
# STFT

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# Results



- Instantaneous amplitude and frequency of two components

# Conclusions

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- The application of statistical model order selection (in this case - estimation of the number of sinusoidal components) allow to track on-line the parameters of the signal.
- The use of information-theoretic criterion like AIC, together with high-resolution parametric estimation method, like MUSIC, allows precise on-line estimation of the signal parameters by using the sliding window approach in the case when the parameters of the components are time-varying.