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ADVCOMP 2021, October 03-07, Barcelona, Spain

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# **Statistical and Principal Component Analysis in the Design of Alkaline Methanol Fuel Cells**

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# Lialia Nikitina

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Dr. Lialia Nikitina, graduated from Moscow State University, MS in 1994, PhD in 2003. Since 2004 she is a research scientist at Fraunhofer Institute for Algorithms and Scientific Computing, Sankt Augustin, Germany. Principal research interests are Mathematical Modeling, Numerical Simulation, Data Analysis, Multidimensional Optimization. She has published more than 50 papers in reputed journals and conference proceedings.



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# Content

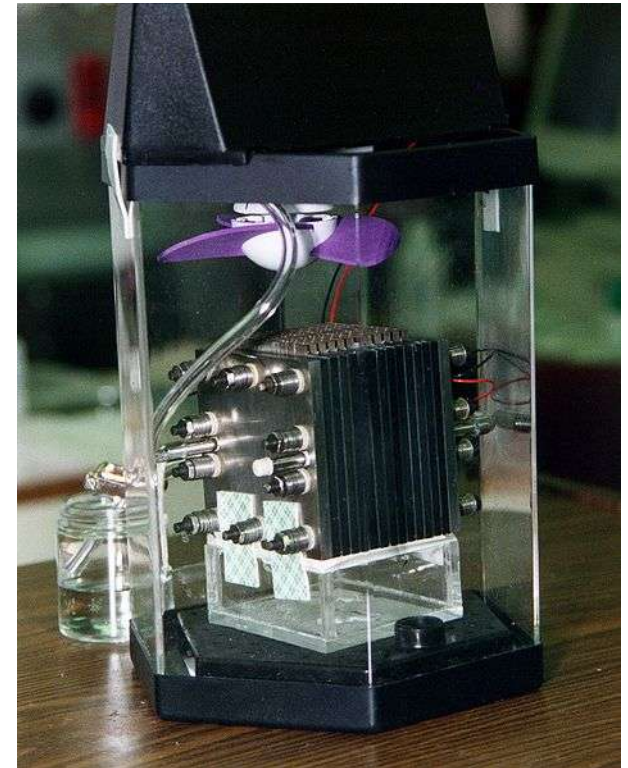
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- Direct methanol fuel cells
- Modeling of electrochemical kinetics
- Fitting of model to experiment
  
- Statistical and Principal Component Analysis
  - sensitivity matrix
  - its singular value decomposition
  - analysis of error propagation

# Electrochemical kinetics of the direct methanol fuel cells

- Fuel cell - similar to galvanic cell but with refilling reagents
- Direct fuel cell - avoids formation of hydrogen

Fuel	Gibbs energy [kJ / mol]	potential [V]	energy density [MJ / kg]
methanol	702	1.213	21.9
ethanol	1325	1.145	28.7
1-propanol	1853	1.067	30.9
2-propanol	1948	1.122	32.4
n-butanol	2602	1.124	35.2
dimethyl ether	1387	1.198	30.3
ethylene glycol	1181	1.224	19.0
formic acid	270	1.400	5.6
glycerol	1622	1.201	16.1



Example (NASA JPL):  
direct methanol fuel cell

# Modeling of electrochemical kinetics

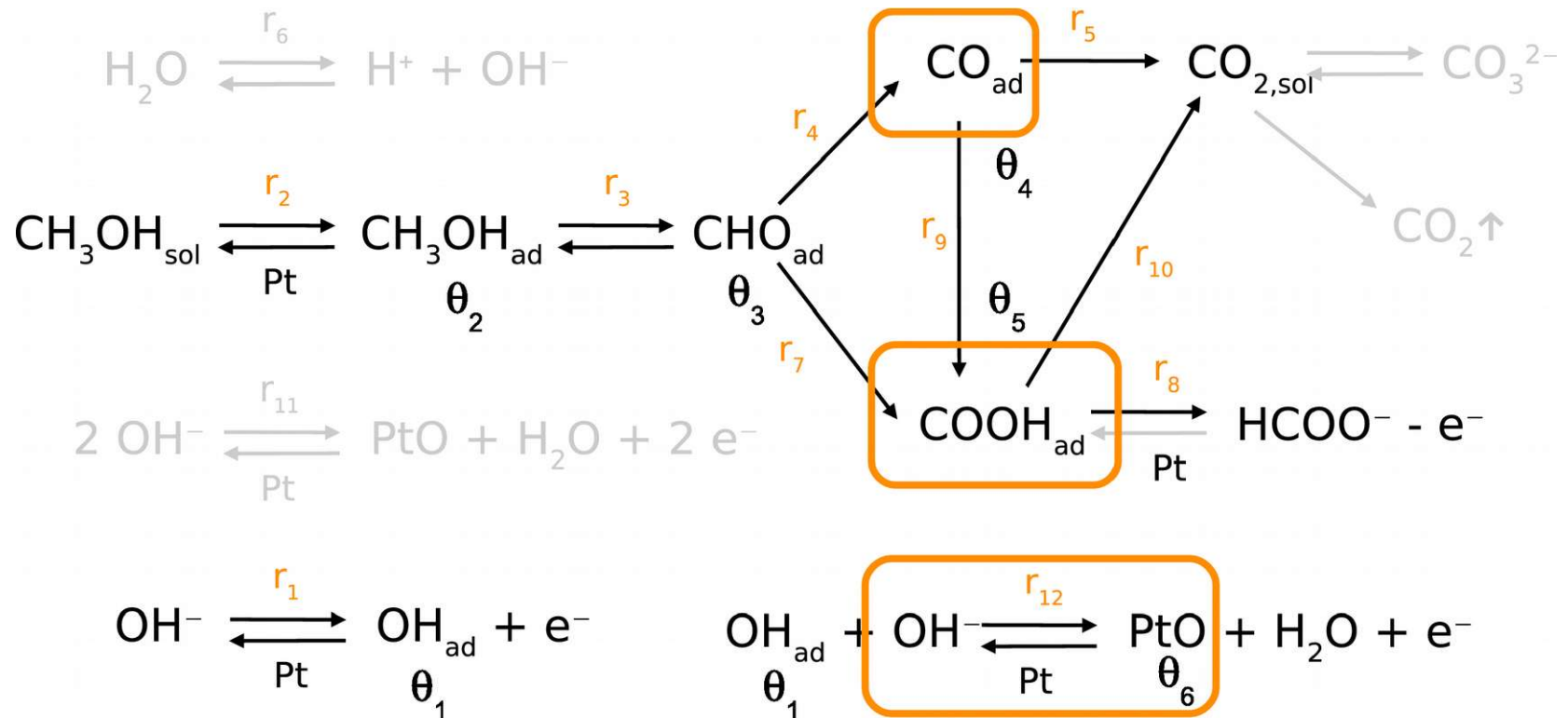
- reactions form a hypergraph

$$reaction_i = \sum_j a_{ij}^L g_j - a_{ij}^R g_j,$$

where  $a_{ij}^L$ ,  $a_{ij}^R$  - incidence matrices multiplied to stoichiometric coefficients, for the left and right hand side of the reaction,  $g_j$ - reagents

- hypergraph is a generalization of graph where an edge can join any number of vertices,  $a_{ij}$  - incidence matrix for ending vertices (j) entering in an edge (i)
- reagents, variables, constants, parameters are listed
- hypergraph is translated to reaction rates  $r_i = k_i^L \prod_j \theta_j^{a_{ij}^L} - k_i^R \prod_j \theta_j^{a_{ij}^R}$
- reaction rates are assembled to molar balance description
- ODEs are formed
- templates for fitting on stationary state and on dynamic system are prepared
- $L_2$ -norm distance between experiments and the model is minimized

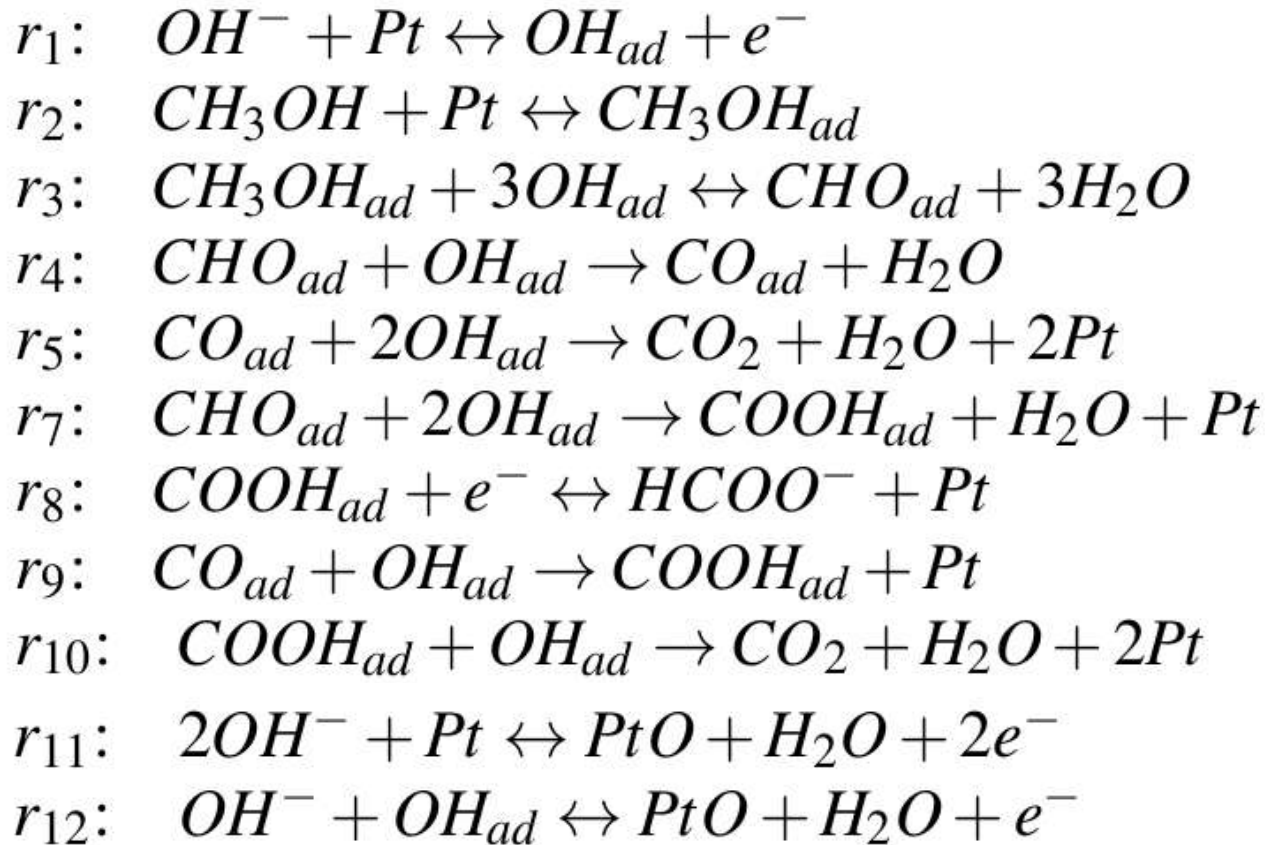
# Mechanism of methanol oxidation: reaction graph



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# Mechanism of methanol oxidation: reactions

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# Mechanism of methanol oxidation: model parameters

model constants

Faraday const  
gas const  
surf.area  
capacity  
activity const  
charge transfer  
coefficient

<i>Constant, units</i>	<i>Value</i>
$F, \text{ C/mol}$	$9.649 \cdot 10^4$
$R, \text{ J/(K mol)}$	8.314
$A, \text{ m}^2$	$2.376 \cdot 10^{-5}$
$C_{dl}, \text{ F}$	$1.899 \cdot 10^{-4}$
$C_{act}, \text{ mol/m}^2$	$8.523 \cdot 10^{-5}$
$\alpha$	0.5

abs.temperature:  $T, \text{ K}$   
voltage:  $\eta, \text{ V}$

<i>Variables</i>		<i>Constants</i>	
$\theta_1$	$OH_{ad}$		
$\theta_2$	$CH_3OH_{ad}$	$c_1$	$OH^-$
$\theta_3$	$CHO_{ad}$	$c_2$	$CH_3OH$
$\theta_4$	$CO_{ad}$	$c_3$	$H_2O$
$\theta_5$	$COOH_{ad}$		
$\theta_6$	$PtO$		

surface  
coverages

volume  
concentrations

current:  $I_{cell}, \text{ A}$



# Mechanism of methanol oxidation: reaction rates

reaction rates  $r_i$ , mol/(m<sup>2</sup>s)

reaction constants  $k_i$ , mol/(m<sup>2</sup>s)

$$r_1 = k_1 c_1 \theta_0 - k_{-1} \theta_1, \quad r_2 = k_2 c_2 \theta_0 - k_{-2} \theta_2,$$

$$r_3 = k_3 \theta_2 \theta_1^3 - k_{-3} \theta_3 c_3^3, \quad r_4 = k_4 \theta_3 \theta_1,$$

$$r_5 = k_5 \theta_4 \theta_1^2, \quad r_7 = k_7 \theta_3 \theta_1^2, \quad r_8 = k_8 \theta_5,$$

$$r_9 = k_9 \theta_4 \theta_1, \quad r_{10} = k_{10} \theta_5 \theta_1,$$

$$r_{11} = k_{11} c_1^2 \theta_0 - k_{-11} c_3 \theta_6,$$

$$r_{12} = k_{12} c_1 \theta_1 - k_{-12} c_3 \theta_6,$$

probability that  
1CO meets 2OH

$$k_1 = k_1^0 \exp(\alpha \beta \eta),$$

$$k_{-1} = k_{-1}^0 \exp(-(1 - \alpha) \beta \eta),$$

$$k_8 = k_8^0 \exp(-(1 - \alpha) \beta \eta), \quad \beta = F / (RT),$$

$$k_{11} = k_{11}^0 \exp(2\alpha \beta \eta), \quad k_{12} = k_{12}^0 \exp(\alpha \beta \eta),$$

$$k_{-11} = k_{-11}^0 \exp(-2(1 - \alpha) \beta \eta),$$

$$k_{-12} = k_{-12}^0 \exp(-(1 - \alpha) \beta \eta)$$

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# Mechanism of methanol oxidation: molar balance

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$$F_1 = (r_1 - 3r_3 - r_4 - 2r_5 - 2r_7 - r_9 - r_{10} - r_{12})/C_{act},$$

$$F_2 = (r_2 - r_3)/C_{act},$$

$$F_3 = (r_3 - r_4 - r_7)/C_{act},$$

$$F_4 = (r_4 - r_5 - r_9)/C_{act},$$

$$F_5 = (r_7 - r_8 + r_9 - r_{10})/C_{act},$$

$$F_6 = (r_{11} + r_{12})/C_{act},$$

$$F_7 = (-r_1 + r_8 - 2r_{11} - r_{12}) \cdot FA/C_{dl}, \quad \longrightarrow \text{electrons come out of here}$$

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# Mechanism of methanol oxidation: problem formulation

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ODEs describing the dynamics

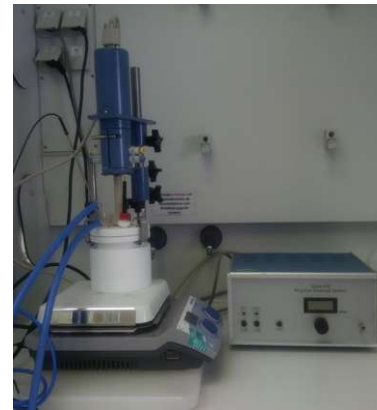
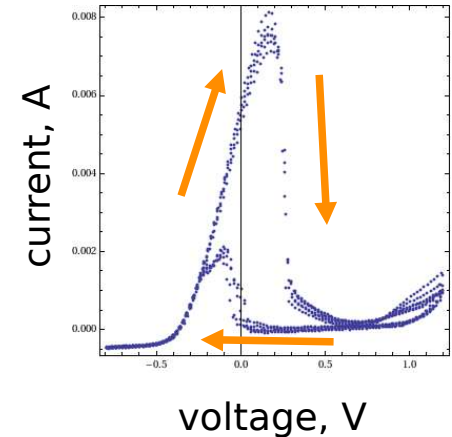
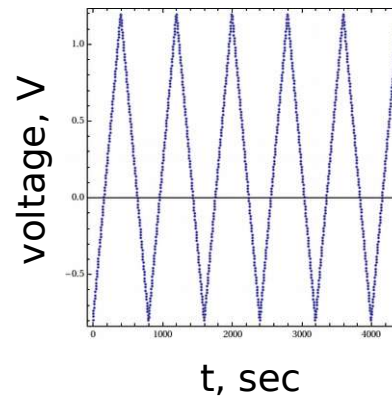
$$d\theta_i/dt = F_i(\theta, \eta), \quad i = 1 \dots n - 1,$$
$$d\eta/dt = F_n(\theta, \eta) + I_{cell}/C_{dl}$$

optimization problem: adjust 14 reaction consts  $k$  to fit experimental data, minimize  $L_2^2 = \sum (I_{cell,i} - I_{cell,i,exp})^2$

stationary state: l.h.s. = 0, algebraic equations

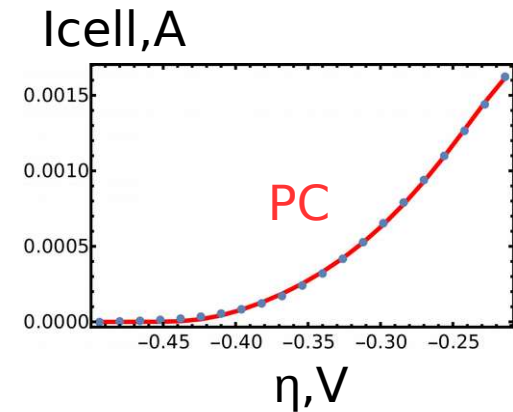
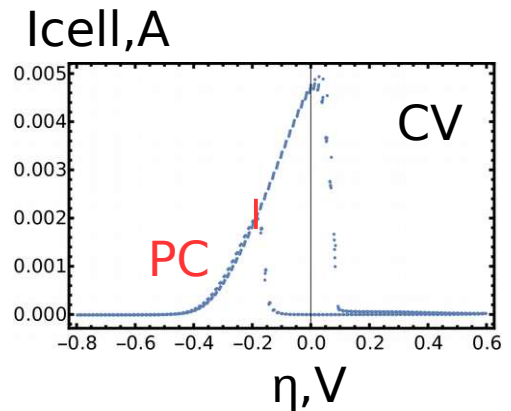
# Fitting of model to experiment

- the cell is probed with saw-like voltage profile
  - high amplitude
  - non-linear effects
  - numerical integration of ODEs
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- teflon cell, deep vacuum to avoid external influence
  - CO<sub>2</sub> is removed by permanent argon blow
  - rotating electrode suppresses diffusion effects



# Fitting of model to experiment

- dynamical part: Cyclic Voltammogram (CV) described by ODE
- stationary part: Polarization Curve (PC) described by algebraic equations, will only be considered in this work
- **blue points** - experimental data
- **red line** - mathematical model
- **perfect fit!**



# Statistical and Principal Component Analysis

TABLE I  
PARAMETER CENTRAL VALUES AND TRUST REGION OF LINEAR MODEL

$p_j$	0.949	-4.5	0.398	-0.563	4.72	-3.46	0.352	-0.101	1.2	-8.66	1.89	-1.08	-1.72	-7.82
$dp_j$	0.3	0.1	0.3	0.4	1.5	0.1	0.1	0.06	0.1	0.1	0.08	0.2	0.15	0.15

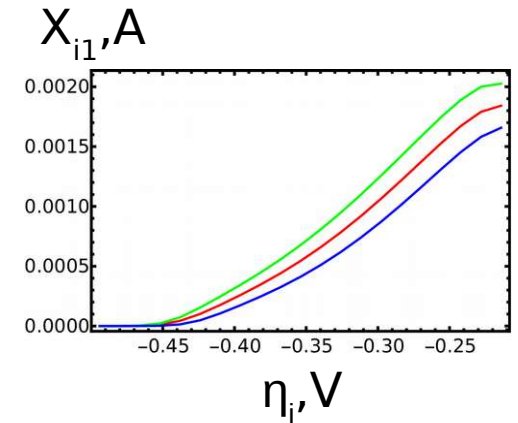
- fitting result (central values,  $p = \log_{10} k$ )
- the main question: **how precise are these values?**
- standard statistical method
- experimental error estimation  $\epsilon^2 = L_2^2 / N_{\text{dof}}$ ,  $N_{\text{dof}} = N_{\text{pt}} - N_{\text{par}}$

$$N_{\text{pt}} = 21, N_{\text{par}} = 14, N_{\text{dof}} = 7,$$
$$L_2^2 = 2.08 \cdot 10^{-9} A^2, \epsilon = 1.72 \cdot 10^{-5} A.$$

# Statistical and Principal Component Analysis

$$X_{ij} = \partial f(\eta_i, p) / \partial p_j, \quad cov = \epsilon^2 (X^T X)^{-1}$$
$$\sigma_i = (cov_{ii})^{1/2}, \quad corr_{ij} = \sigma_i^{-1} cov_{ij} \sigma_j^{-1},$$

- sensitivity, covariance, correlation matrices
- diagonal of cov: squared errors of parameters
- the deviation of the curve (~20%) measures non-linearity of the model inside given dp-box
- surprise!  $X^T X$  is degenerate, cannot be inverted
- a deeper analysis is needed



sensitivity matrix  
(the first column)  
**red** - central diff  
scheme,  
**green/blue** -  
forward/backward  
diff schemes

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# Statistical and Principal Component Analysis

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- trust region: box in parameter space, marking the applicability of linear model up to a given tolerance (e.g., 20%)
- error ellipsoid: one-sigma confidence region in parameter space, corresponding to the estimated Gaussian errors of experiment
- semi-axes of the ellipsoid that belong to trust region, correspond to parameter combinations, which can be measured precisely
- other directions are measured imprecisely

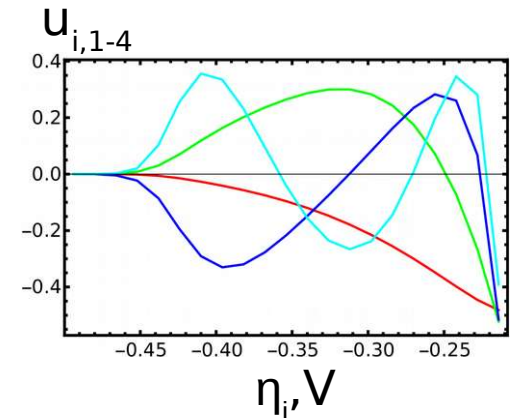
$$X = u\lambda v^T, \quad u^T u = 1, \quad v^T v = v v^T = 1, \quad a_k = \epsilon / \lambda_k.$$

- Singular Value Decomposition (SVD):  
 $X$  is  $N_{pt} \times N_{par}$  rectangular,  $u$  is  $N_{pt} \times N_{par}$  semi-orthogonal,  
 $\lambda$  is  $N_{par} \times N_{par}$  diagonal,  $v$  is  $N_{par} \times N_{par}$  orthogonal,  
 $a_k$  values represent the semi-axes of the error ellipsoid



# Statistical and Principal Component Analysis

- cont'd: columns of v-matrix represent directions of the axes of the ellipsoid in parameter space
- columns of u-matrix represent the profiles of principal components in the space of experiments
- these profiles show the variation of PC curve when the parameters are displaced along the axes of the error ellipsoid



u-profiles for the first four components, in **red-green-blue-cyan** order for  $u_{i,1-4}$

# Statistical and Principal Component Analysis

- the main result: 4 directions are measured precisely, 9 imprecisely, 1 cannot be measured in principle (exact symmetry of the system)

TABLE II  
RESULTS OF PRINCIPAL COMPONENT ANALYSIS

$\lambda_k, A$	$a_k$	$v_{jk}$														
$8.12 \cdot 10^{-3}$	$2.12 \cdot 10^{-3}$	-0.531	0.014	0.461	-0.459	0.140	-0.013	-0.007	0	0	0	0	-0.522	0.032	-0.032	
$1.21 \cdot 10^{-3}$	$1.43 \cdot 10^{-2}$	0.396	-0.159	-0.251	0.250	0.214	-0.010	-0.018	0	0	0.002	0	-0.769	0.167	-0.167	
$2.71 \cdot 10^{-4}$	$6.36 \cdot 10^{-2}$	-0.078	0.639	-0.052	0.058	-0.402	0.065	-0.065	0	0	0.004	0	-0.052	0.451	-0.451	
$1.24 \cdot 10^{-4}$	0.139	-0.066	-0.711	0.091	-0.074	-0.191	-0.058	0.066	0	0	-0.014	0	0.197	0.443	-0.443	
$7.79 \cdot 10^{-6}$	2.21	0.305	-0.063	0.211	-0.123	-0.728	-0.290	0.277	0	-0.001	-0.113	0	-0.234	-0.206	0.206	
$5.1 \cdot 10^{-6}$	3.38	0.546	0.214	0.540	-0.152	0.408	-0.204	0.205	0	0	-0.056	0	0.192	0.160	-0.160	
$8.51 \cdot 10^{-7}$	$2.02 \cdot 10^1$	0.315	-0.066	0.148	-0.264	-0.163	0.336	-0.449	0.001	0.032	0.679	-0.003	-0.010	-0.033	0.033	
$1.69 \cdot 10^{-7}$	$1.02 \cdot 10^2$	0.033	-0.083	0.508	0.459	-0.114	0.441	-0.349	-0.001	-0.029	-0.434	0.003	-0.033	-0.046	0.046	
$2.52 \cdot 10^{-8}$	$6.84 \cdot 10^2$	0.240	-0.003	-0.304	-0.617	-0.003	0.439	-0.030	-0.003	-0.080	-0.518	0.012	0.021	0.013	-0.013	
$7.81 \cdot 10^{-9}$	$2.21 \cdot 10^3$	0.047	-0.001	-0.075	-0.135	0.004	-0.582	-0.722	0.008	0.212	-0.257	-0.039	0.031	0.004	-0.004	
$5.92 \cdot 10^{-10}$	$2.91 \cdot 10^4$	0	0	-0.002	-0.002	0	-0.165	-0.159	-0.037	-0.972	0.022	0.002	0.007	0	0	
$9.35 \cdot 10^{-13}$	$1.84 \cdot 10^7$	0	0	0	0	0	0.024	0.024	0.485	-0.028	0	-0.874	0	0	0	
$3.38 \cdot 10^{-13}$	$5.09 \cdot 10^7$	0	0	0	0	0	0.014	0.014	-0.874	0.028	0	-0.485	0	0	0	
0	$\infty$	0	0	0	0	0	0	0	0	0	0	0	0	0	0.707	0.707

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# Statistical and Principal Component Analysis

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- exact symmetry of the stationary system
- revealed as a result of PCA
- variation  $\delta p_{13} = \delta p_{14}$  corresponds to infinite scatter
- the result depends only on  $p_{13}-p_{14}$ , or on ratio of corresponding k's
- the source of degeneracy of  $X^T X$  matrix
  
- Implementation in *Mathematica*
- `Nsolve` - to solve stationary algebraic system, real roots selected
- `Manipulate` - interactive change of parameters to find starting point
- `FindMinimum` / `Nminimize` - for local/global optimization
- `NonlinearModelFit` - fitting interface to the methods above
- `Gradient` - option to the fitting method to provide finite diff scheme
- `SingularValueDecomposition` - SVD/PCA

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# Conclusion

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- the electrochemical alkaline methanol oxidation process is mathematically modeled
- an algorithm for reconstructing the reaction constants from the experimentally measured polarization curve is developed
- the approach combines statistical and principal component analysis
- formal criteria for reconstruction accuracy based on the estimate of the trust region for the linearized model are defined

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# Conclusion

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- analysis shows that the described experiment allows to determine precisely not all 14 reaction constants, but only 4 their certain linear combinations
- of the remaining orthogonal combinations, one corresponds to the symmetry of the stationary system and is fundamentally indeterminate in the described experiment
- the remaining 9 combinations have insufficient reconstruction accuracy
  
- Further plans: other experiments should be involved in the analysis, including fully dynamic cyclic voltammetry and variations in the concentration of the main reagents.