

# Achievements and Challenges in Environmental Sensing

## Moderator:

Dr. Sergey Y. Yurish, IFSA President, Spain

## Panelists:

**Winfried Vonau**, Kurt Schwabe Research Institute, Germany

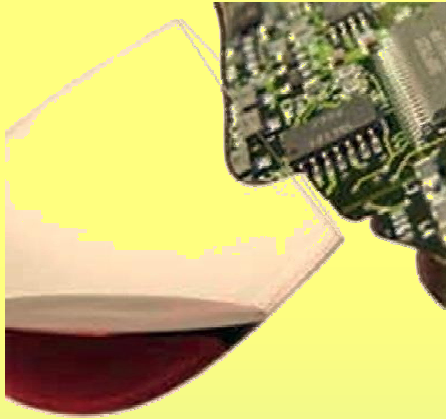
**Ilia Kiselev**, Breitmeier Messtechnik GmbH, Germany

**Andrey Teplykh**, Kotel'nikov Institute of Radio Engineering  
and Electronics of RAS, Russia



# Panel Conclusions

- Main challenges in environmental sensing are molecular sensors
- Namely such sensors let to achieve low cost and low power consumption in various applications, including IoT (smart cities, etc.)
- Progress in sensor interfacing and readout is connected with a changing of traditional sensor output's informative parameters as voltage and current) to frequency-time informative parameters



Is it possible to bring metal oxide E-noses to every-day application, in environmental measurements?

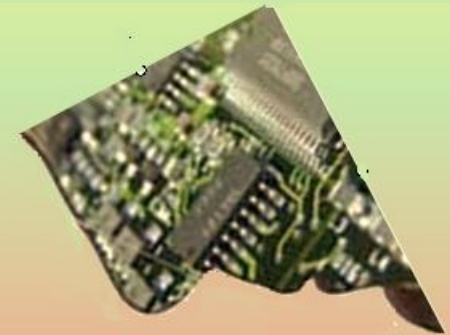
***Ilia Kiselev***

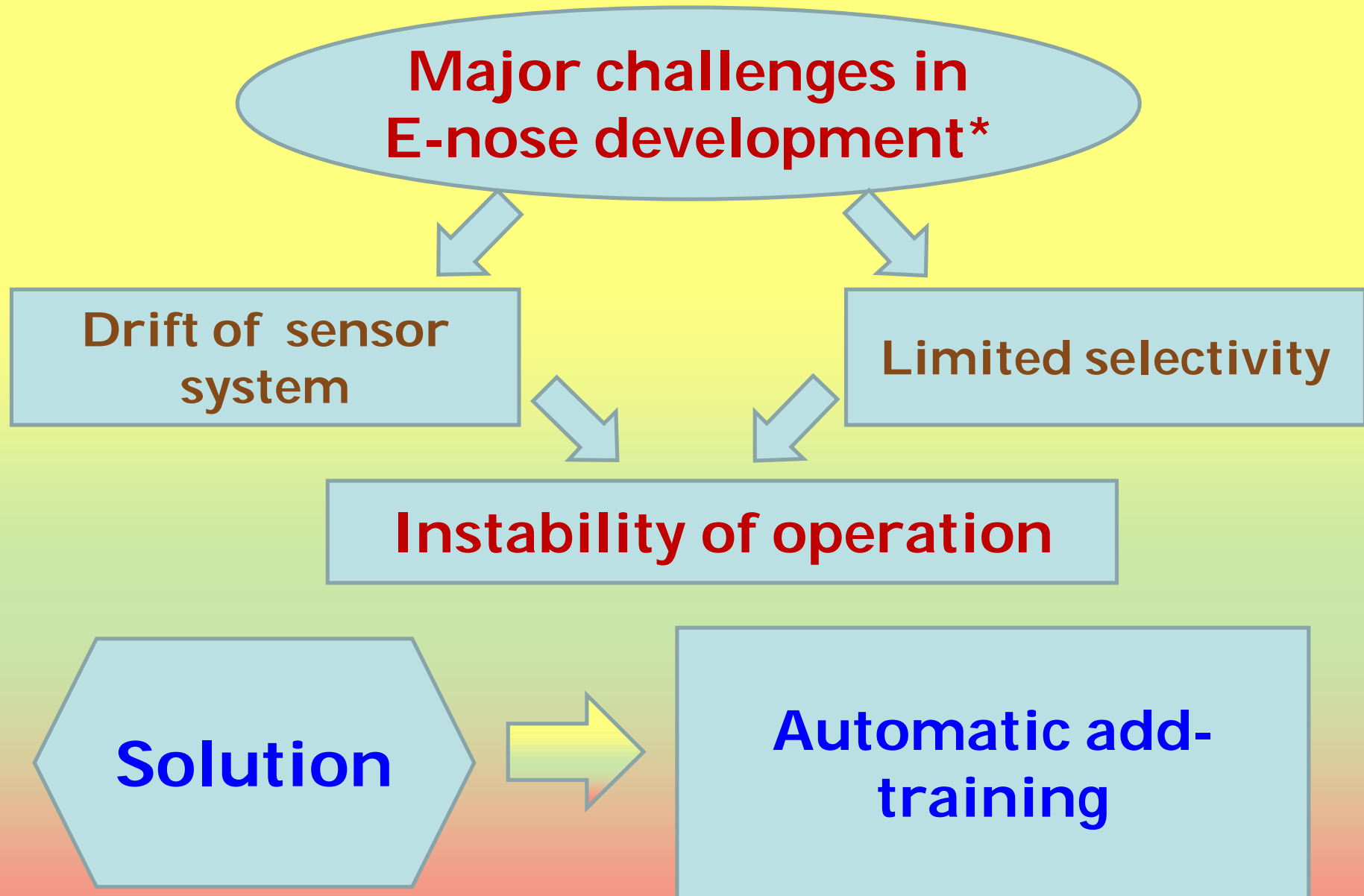
Breitmeier Messtechnik GmbH, Germany  
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**Major perception of  
E-noses appearance**



**There is no working  
devices of such a type  
around**





\* A. Loutfi, S. Coradeschi, G. K. Mani, P. Shankar, J. B. B. Rayappan, *Electronic Noses for Food Quality: A Review*, Journal of Food Engineering, 144, 2015.

# Add-training is indispensable

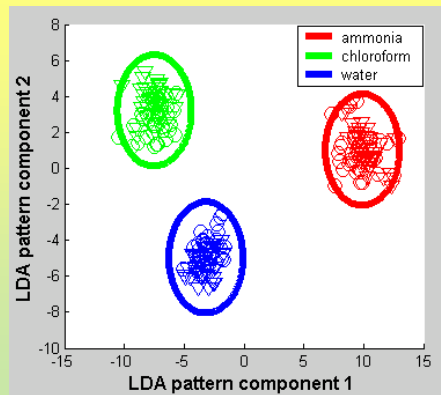
Three time scales of the odor recognition system change in ambient air:

1. **Primary variations** (concentration, daytime air, etc. variations): **1 – 2 days**;
2. **Regular variations** (object change, week day air variations, etc.): **1-2 weeks**;
3. **Drift** (season change, sensor ageing): **months, years**.

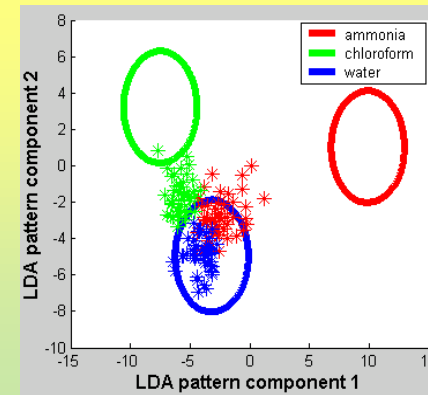
Three stages of the model add-training

1. **Initial model building** (many expositions a day);
2. **Model completion** (few expositions a day);
3. **Continues add-training** (dropping exposition frequency, from one a day to one per couple of months).

Initial model building: separation is perfect

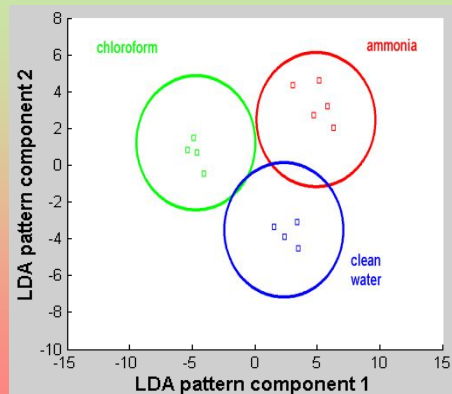


Further exposures slip out

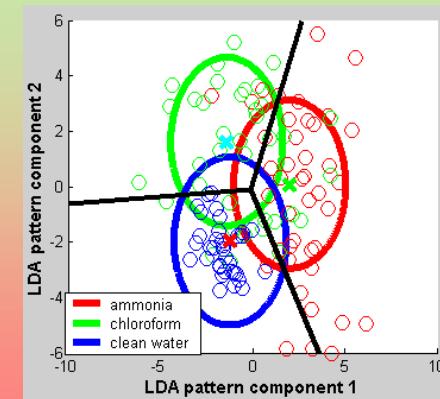


Are taken into training set

Work model: separation is optimal

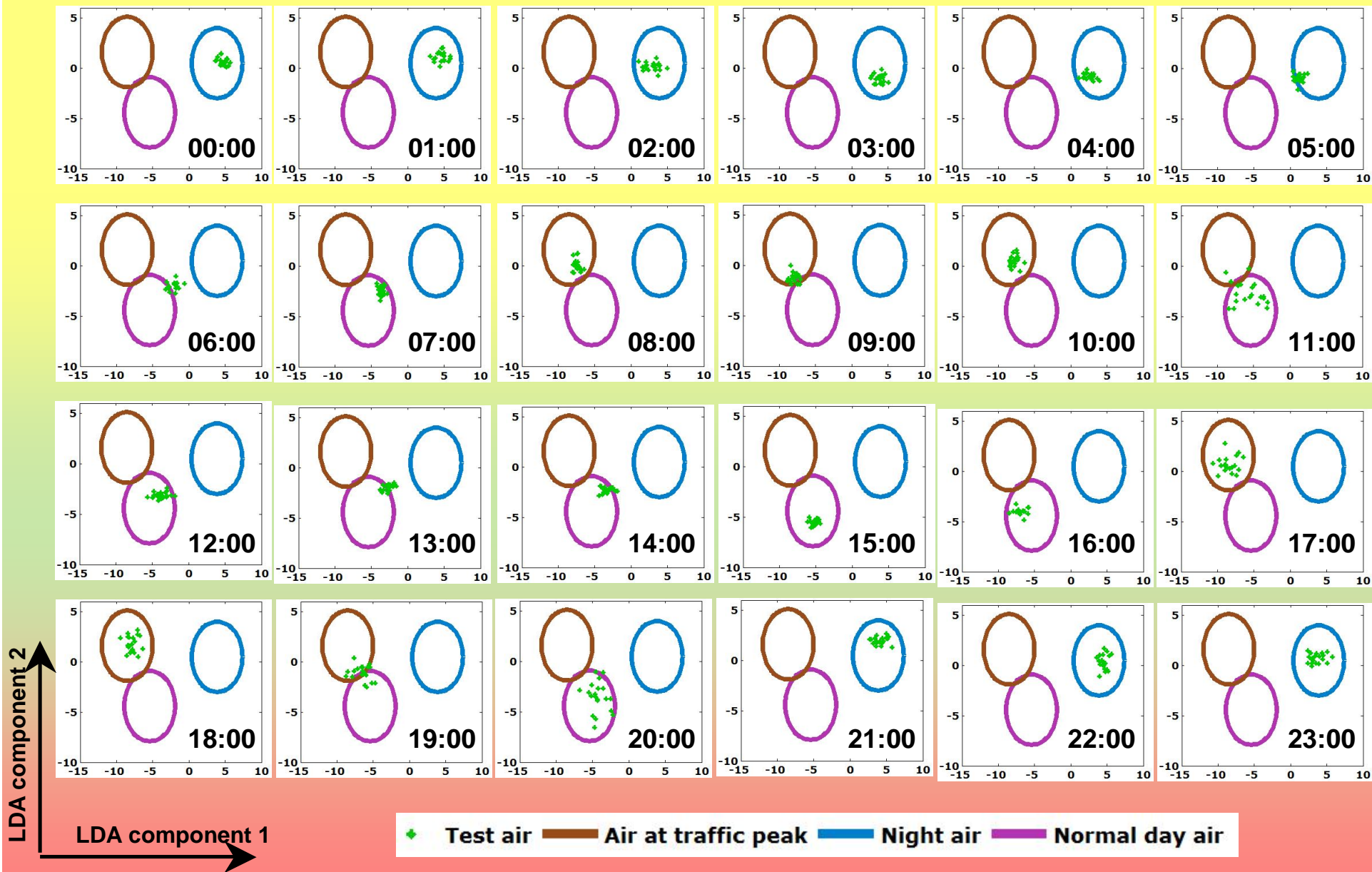


Aging: separation becomes insufficient



Withdrawal of obsolete training data

# Example: Environmental air monitoring with metal oxide sensor array following proper training





## SENSORDEVICES 2016



**Example: A MO gas sensor array operates for 15 years in a continuous mode to demonstrate 100% recognition of smells of lighter, felt-tip pen and ambient air. Add-training: one exposure every 2 months.**





# Conclusions

1. The future and broad usage belong to molecular-group-specific noses similar to the biological ones.
2. The metal oxide E-noses, as other non-molecular-specific noses, have their niche for application.
3. **It is already certain that the existing E-noses can be used for a “simple” long-term application: tracing of expressed pollutions, discrimination of distant odors. This is achieved using the add-training.**
4. **It is still the main challenge: to get really working industrial E-noses for such simple applications.**
5. MO E-noses won't take their place in the market without demonstrative, long working really useful applications. The letter should be established using unprofitable fund raising.
6. It is known that MO E-noses can discriminate very slightly differing odors, such as required for classification e. g. of wine sorts. But it still must be proven, whether long-term applications of this kind are feasible – specifically, using the add-training.



# **Achievements and Challenges in Environmental Sensing**

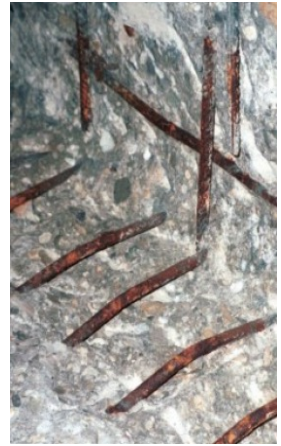
Electrochemical sensors

Winfried Vonau  
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04736 Waldheim, Germany

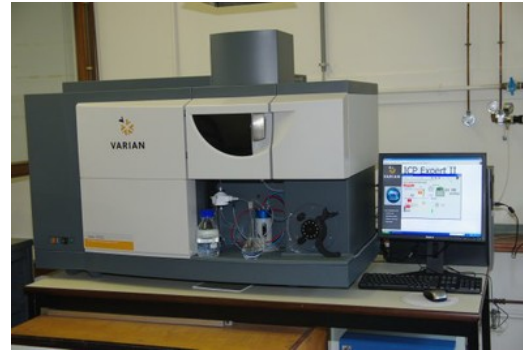
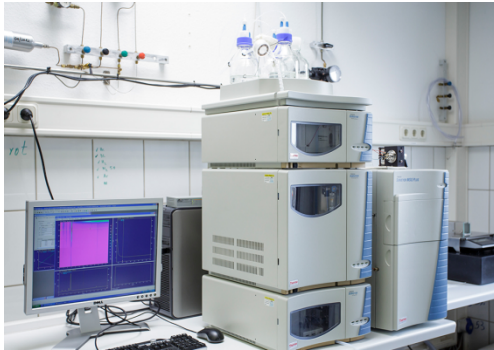
# Environmental analysis with (electro-) chemical sensors I



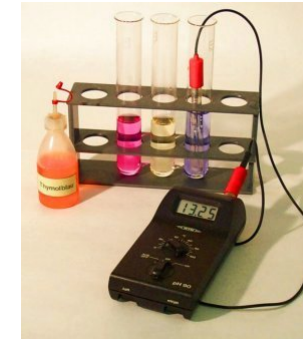
# Environmental analysis with (electro-) chemical sensors II



## Expensive laboratory analysis using costly equipment



## On-site measurements using sensors



## Requirements on [(electro)chemical] sensors as an alternative to other analytical tools

- \* High selectivity and sensitivity
- \* Low detection limit
- \* Low signal noise

- \* no hysteresis
- \* short response times
- \* long life-time

- \* sterilisability
- \* inexpensive production
- \* ...

## Requirements for environmental sensing (electrochemical sensors):

### Detection possibilities

No sensors for sulfate and phosphate and other environmentaloly relevant species available

### Detection limits

Sensors must become better (exception: pH glass electrode)

### Cross sensitivity

Often much too high

### Long-term stability

Often too much calibration cycles necessary

### Multi component analysis

Often too much calibration cycles necessary

...

## Miniaturisation

All solid state sensors

Planar sensors

...



# Applications of Acoustic Resonance Spectroscopy for Sensor Devices

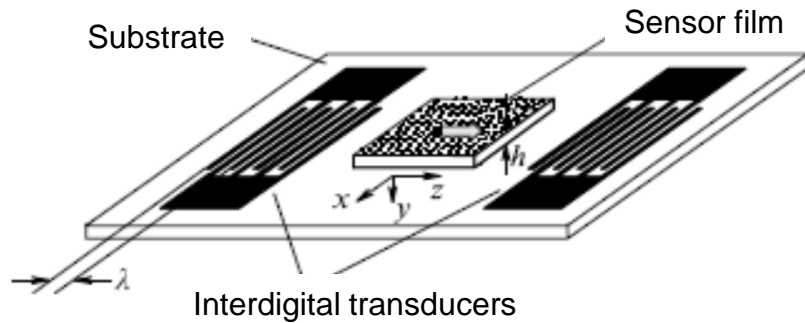
Andrey Teplykh

Kotel'nikov Institute of Radio Engineering and Electronics of RAS, Saratov Branch

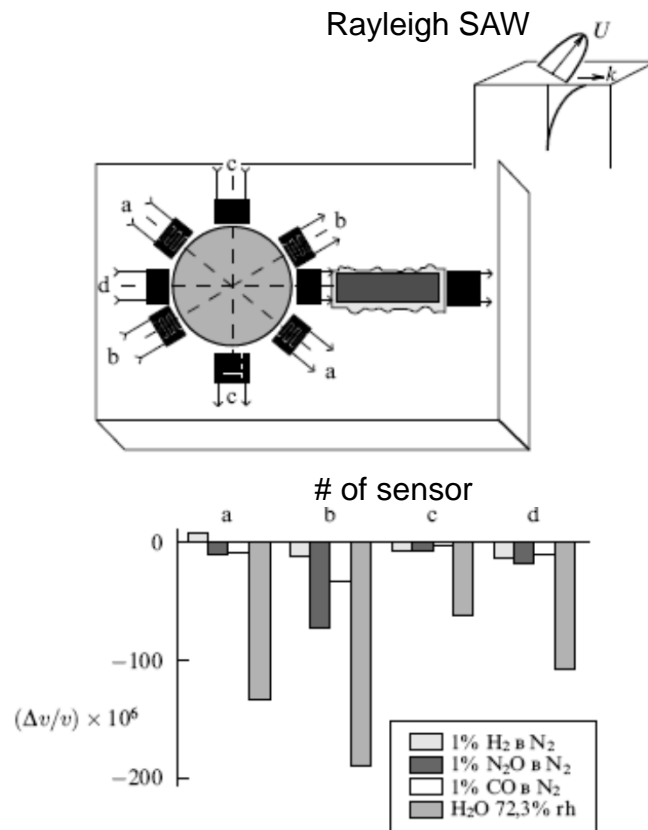
E-mail: `teplykhaa@mail.ru`

SENSORDEVICES 2016, July 24 - 28, 2016, Nice, France

# Gas acoustic sensors



SAW based single channel gas sensor  
**sensitive to surface massloading**

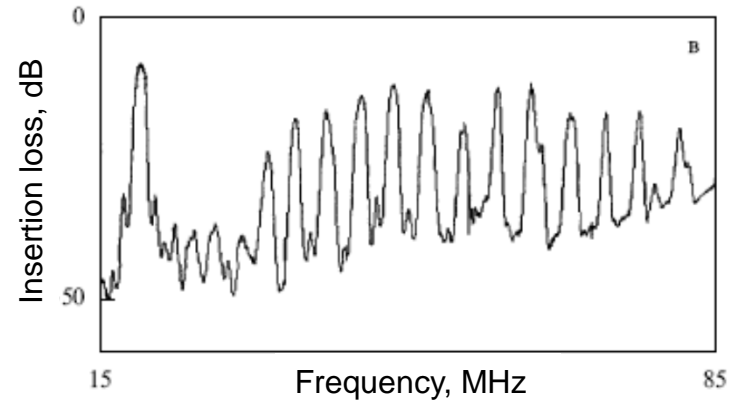
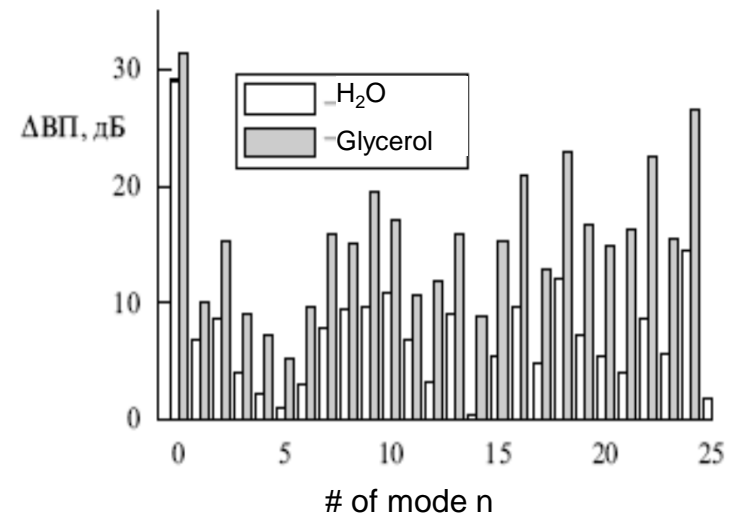
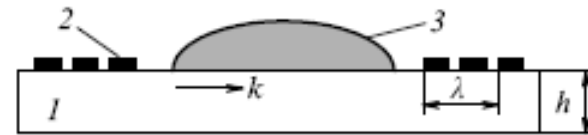
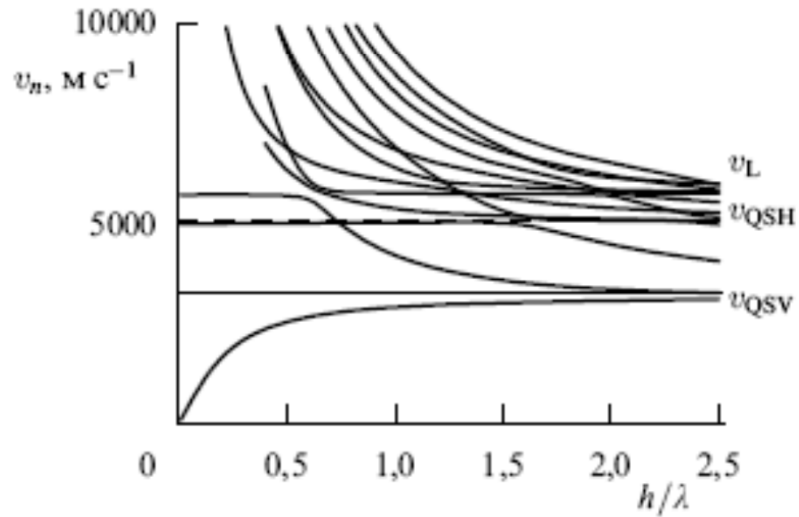
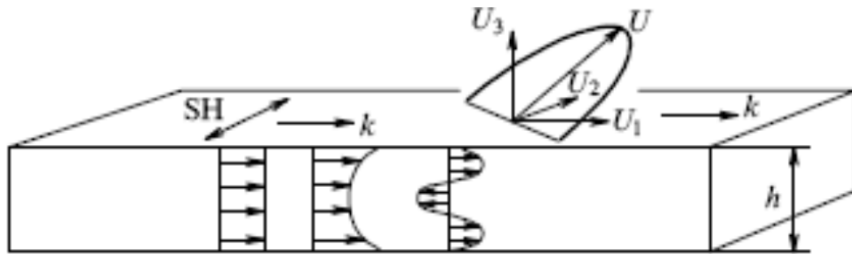


Integral array of SAW sensors  
 exploit substrate anisotropy  
**sensitive to different gas absorption**

V. I. Anisimkin *et al.*

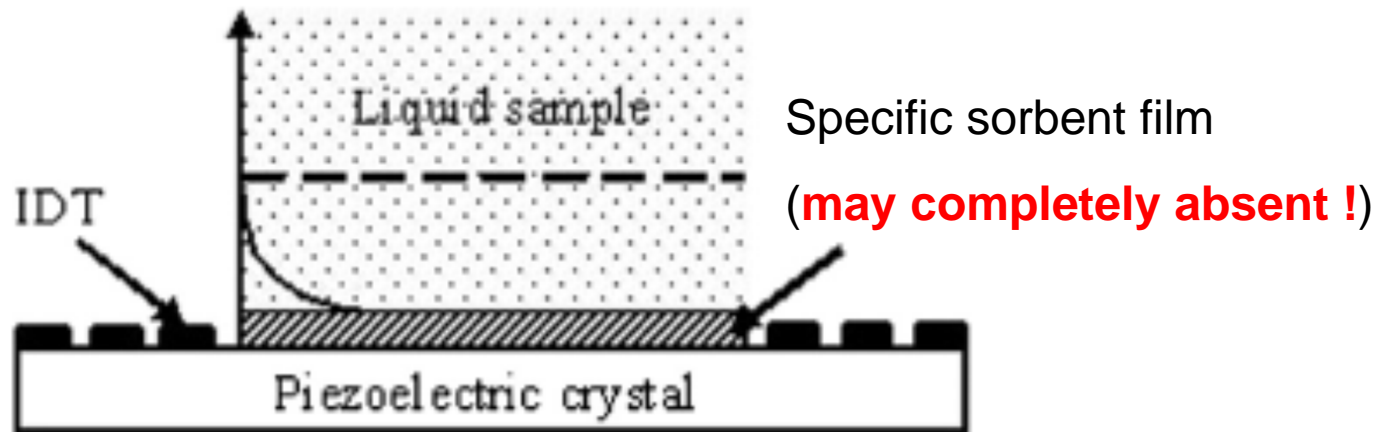
# Liquid acoustic sensors

Plate acoustic waves (different modes)



V. I. Anisimkin *et al.*

# Bacterial cell acoustic sensors



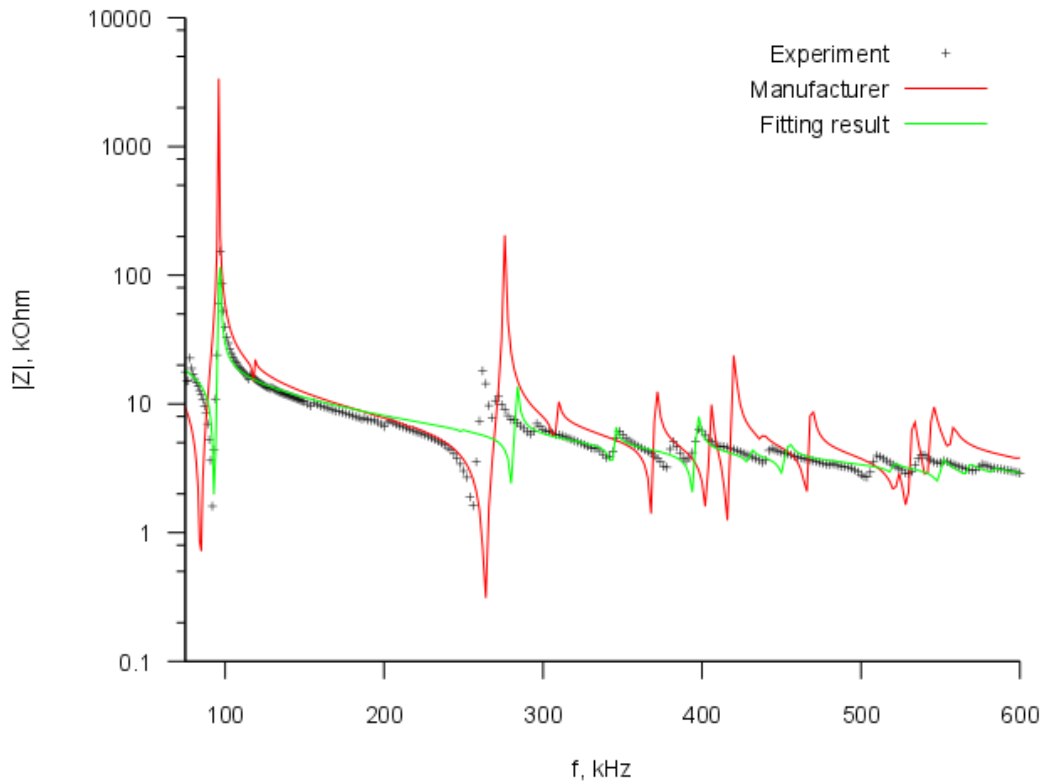
Schematic view of acoustic wave liquid sensor

Ultrasonic acoustic waves propagating in thin piezoelectric plates with free faces are used for bacteria detection in micro-litre liquid samples deposited on one of the plate surface. The limits of the detection at normal conditions are as low as 0.04% for highly diluted rich cultural Luria–Bertani broth (LB-media) in distillate water, 0.07% for bacterial cells in distillate water, and 0.6% for bacterial cells in LB-media.

**For all analytes the most probable detection mechanism is the change in liquid conductivity.**

V. V. Kolesov *et al.*

# Identification of acoustic properties of materials: acoustic spectroscopy approach



$$\begin{pmatrix} S_1 \\ S_2 \\ S_3 \\ S_4 \\ S_5 \\ S_6 \\ D_1 \\ D_2 \\ D_3 \end{pmatrix} = \begin{pmatrix} s_{11}^E & s_{12}^E & s_{13}^E & 0 & 0 & 0 & 0 & 0 & d_{31} \\ s_{12}^E & s_{11}^E & s_{13}^E & 0 & 0 & 0 & 0 & 0 & d_{31} \\ s_{13}^E & s_{13}^E & s_{33}^E & 0 & 0 & 0 & 0 & 0 & d_{33} \\ 0 & 0 & 0 & s_{55}^E & 0 & 0 & 0 & d_{15} & 0 \\ 0 & 0 & 0 & 0 & s_{55}^E & 0 & d_{15} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & s_{66}^E & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & d_{15} & 0 & \epsilon_{11}^T & 0 & 0 \\ 0 & 0 & 0 & d_{15} & 0 & 0 & 0 & \epsilon_{11}^T & 0 \\ d_{31} & d_{31} & d_{33} & 0 & 0 & 0 & 0 & 0 & \epsilon_{33}^T \end{pmatrix} \times \begin{pmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \\ T_5 \\ T_6 \\ E_1 \\ E_2 \\ E_3 \end{pmatrix}$$

**Clarification of material constants sets allows more accurately modeling the newly created sensors**