

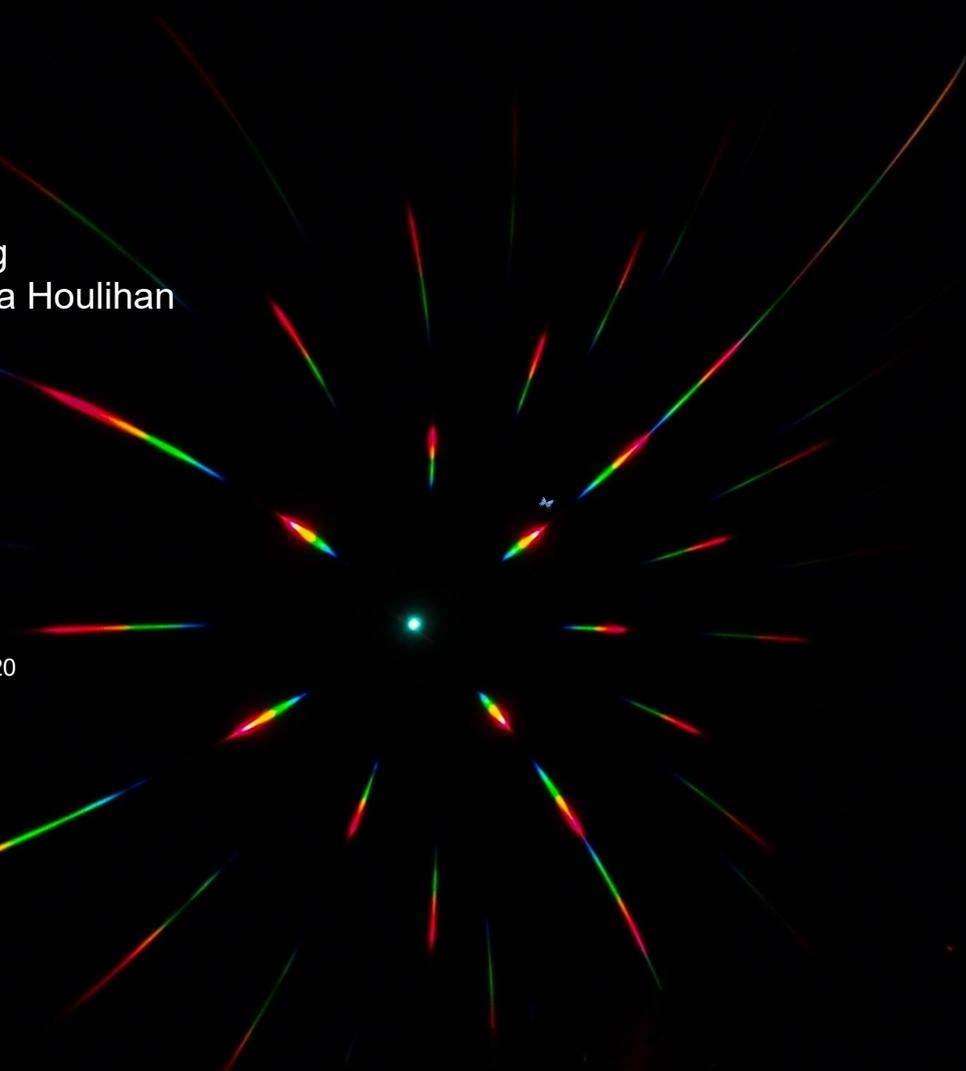
Multi-Gas Sensors for Enhanced Reliability of SOFC Operation

DOE/NETL Cooperative Agreement: DE-FE0031653

GE: Radislav Potyrailo, Joleyn Brewer, Baokai Cheng
SUNY Polytechnic Institute: Michael Carpenter, Nora Houlihan

NETL Program manager: Venkat K. Venkataraman

21st Annual Solid Oxide Fuel Cell (SOFC) Project Review Meeting, July 21-23, 2020



Project goal and objectives

Goal:

to build gas sensors for in situ monitoring of H₂ and CO gases of SOFC systems

Objectives:

to achieve multi-gas monitoring capability with a single multivariable sensor and to achieve its long-term sensor performance

Real-time knowledge of H₂/CO ratio of anode tail gases:

- will allow control of efficiency of reforming process in the SOFC system
- will deliver a lower operating cost for SOFC customers

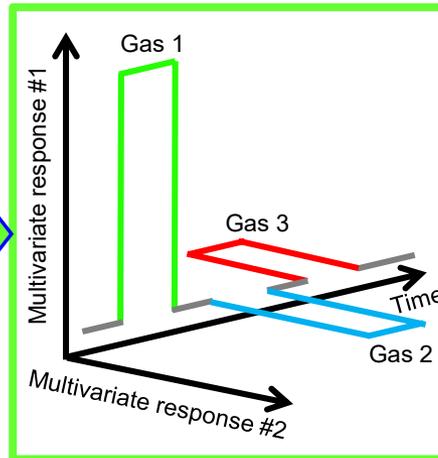
Status quo:

Mature traditional detector concepts



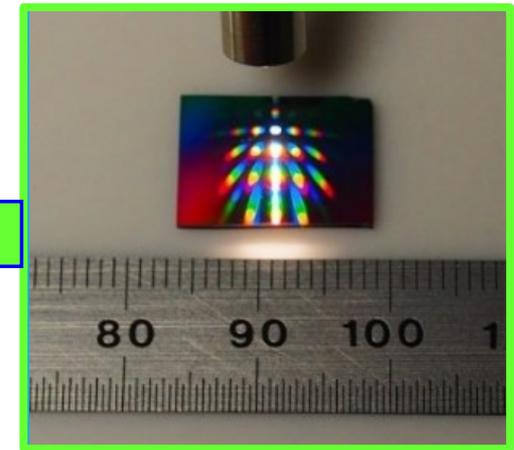
Performance need:

Compact system for multi-gas discrimination



Our approach:

Multivariable photonic gas sensors



In-line sensing is not straightforward, requires traditional analytical instruments

Modern conventional gas sensors:

“For the revolution to take off, accuracy must improve”

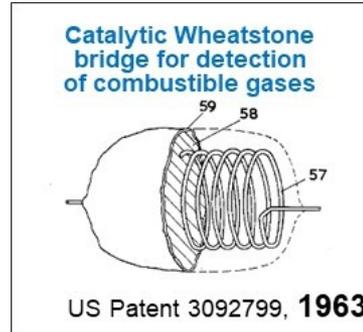


- ✓ Miniaturization
- ✓ Reduced power
- ✓ Low cost

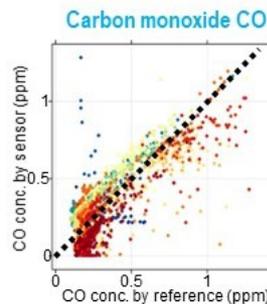
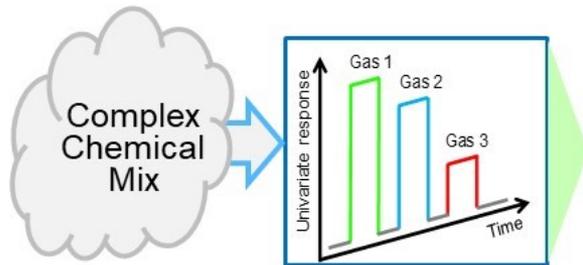
alphasense.com
figaro.com
sgxsensortech.com

The roots: industrial safety

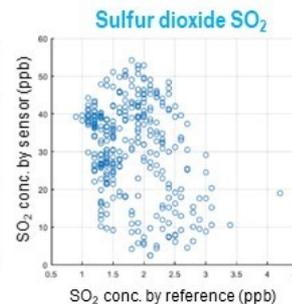
single-output devices for detection of expected gases at high concentrations



Gas cross-sensitivity: undesired response to interfering gases



Hagler et al., Atmospheric Meas. Tech. 2016, 9, 5281



Borrego et al., Atmospheric Environment 2016, 147, 246



No low cost sensors meet the Regulatory Monitoring requirements.
-Air Sensor Guidebook, EPA/600/R-14/159, 2014

nature

The biggest headaches are caused by interfering chemicals.
-Lewis, Edwards, 2016

High performance analytical instruments: Diverse designs to reject known and unknown interferences

Gas chromatography



Photo by R. Potyrailo

Qin, Gianchandani, *Microsyst. Nanoeng.* 2016

Mass spectrometry



Photo by R. Potyrailo

May 28, 2018

Laser spectroscopy



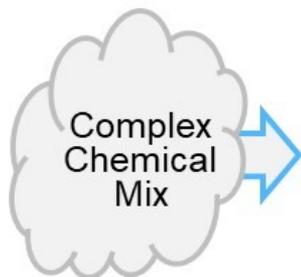
www.eol.ucar.edu

alliedscientificpro.com

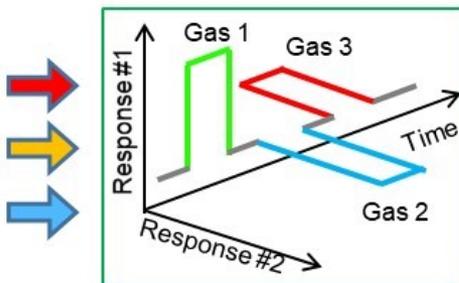
Multi-detectors



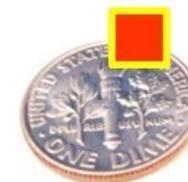
emersonprocess.com



<http://www.yalescientific.org>



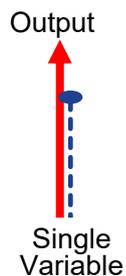
Traditional analytical instrument



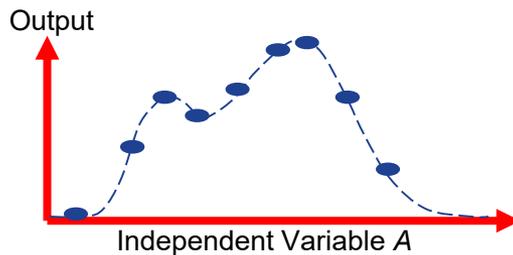
Diverse instrument-design rules to operate in complex conditions

Design principles of analytical instruments: Different orders of measurements

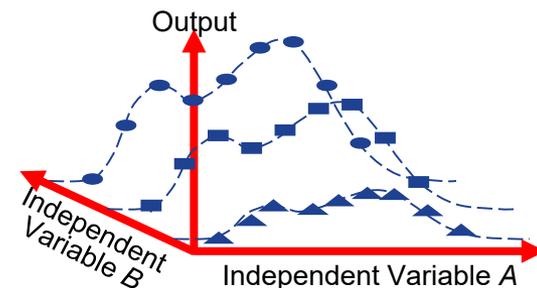
Zero order



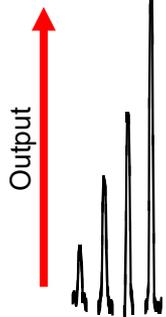
First order



Second order

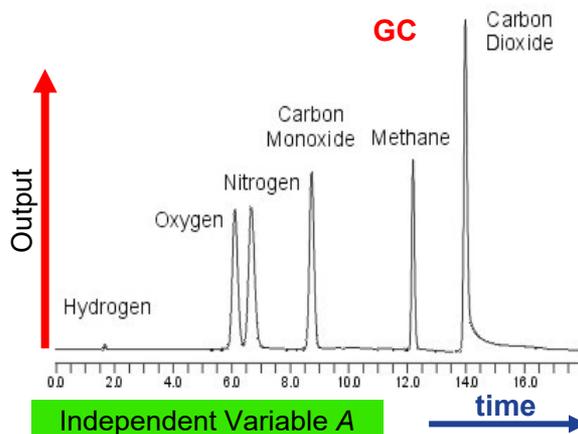


Zero order



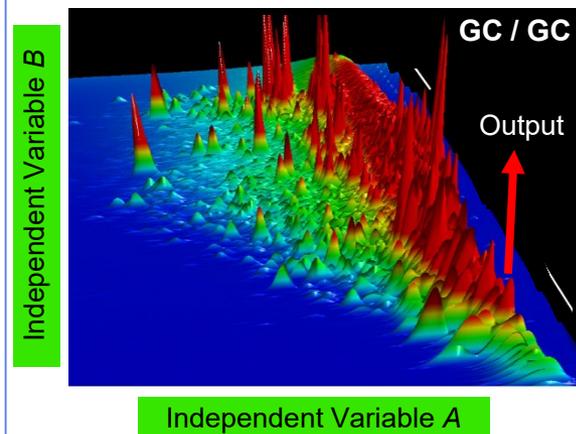
No independent variables

First order



One independent variable allows rejection of KNOWN interferences

Second order



Two or more independent variables reject UNKNOWN interferences

Mathematically, interferences are not noticed, cannot be rejected

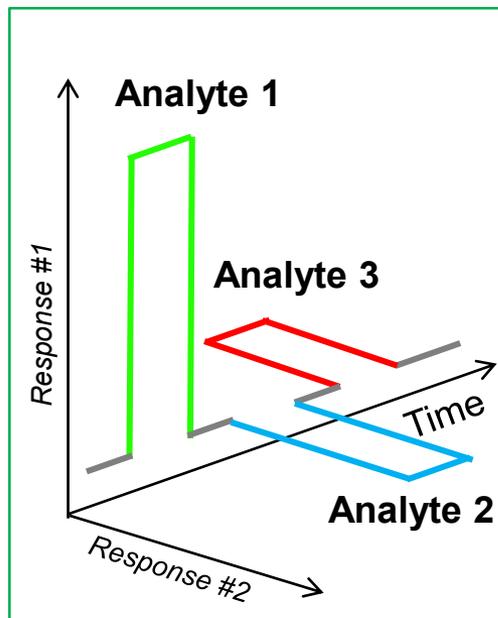
GE Research:

multivariable sensor solutions for demanding applications

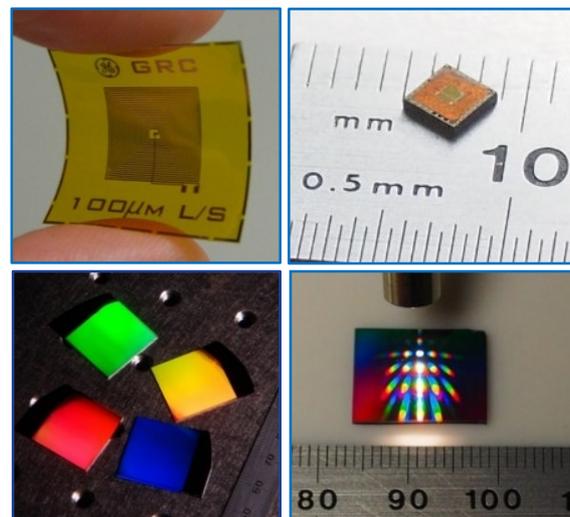
Instruments based on mature analytical concepts



Qin, Gianchandani, *Microsyst. Nanoeng.* 2016
alliedscientificpro.com
emersonprocess.com
C&EN May 28, 2018



Sensors and systems based on multivariable sensing concepts



Photos by R. A. Potyrailo

Bringing contemporary sensing solutions to society using mathematics and physics

Our industrial R&D goal:

Develop sensors with new capabilities, transition for commercialization

Technology effort

Early TRL research



GE Research

Commercialization

WIRELESS MULTIPARAMETER GAS SENSOR NETWORKS



2019

RF MULTIPARAMETER OIL SENSOR



2017

GE Ventures START-UP RF MULTI-GAS SENSORS



2014



2011

MULTI-ION WATER SENSORS



GE Businesses, External partners



TRL = technology readiness level
NPI = new product introduction



Our industrial R&D goal:

Develop sensors with new capabilities, transition for commercialization

Technology effort

Early TRL research



GE Research

Commercialization



WIRELESS MULTIPARAMETER GAS SENSOR NETWORKS

2019



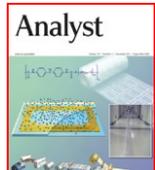
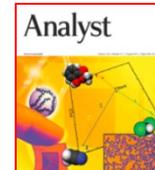
RF MULTIPARAMETER OIL SENSOR

2017



GE Ventures START-UP RF MULTI-GAS SENSORS

2014



GE Businesses, External partners



MULTI-ION WATER SENSORS

2011



Idea/discovery

Feasibility

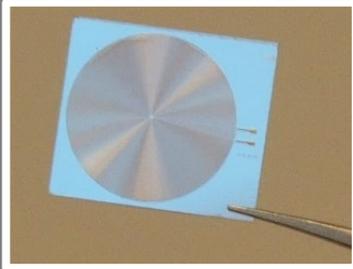
Tech transfer

NPI

Product maturity

TRL = technology readiness level
NPI = new product introduction

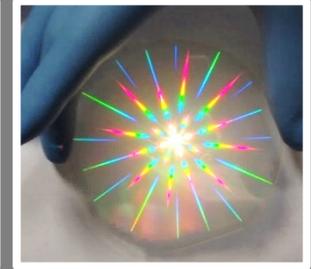
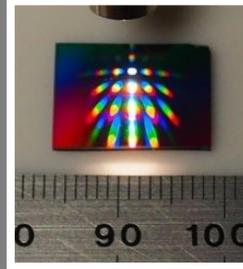
Our design principles for multivariable gas sensors



Radio frequencies



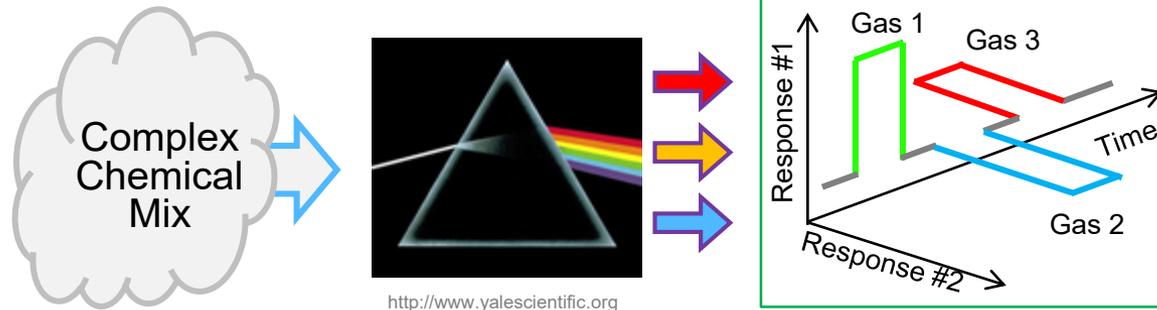
Microwave frequencies



Optical frequencies

Potyrailo et al., *Nature Electronics* 2020

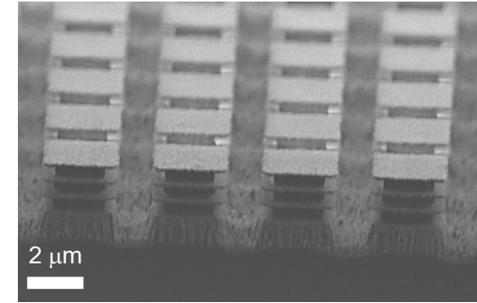
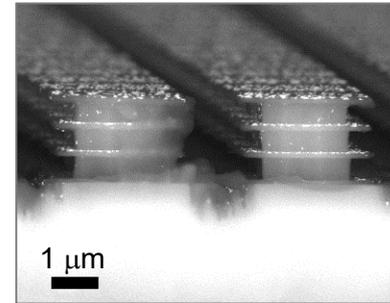
Potyrailo et al., *Nature Photonics* 2007
Potyrailo et al., *Proc. Natl. Acad. Sci. USA* 2013
Potyrailo et al., *Nature Communications* 2015



<http://www.yalescientific.org>

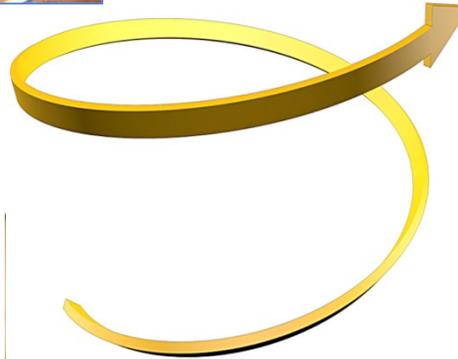
New philosophy for selective gas sensing across different frequency ranges

Learning from Nature



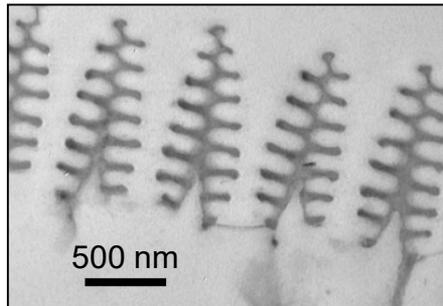
Bioinspiration -
new functionality, beyond Nature
High temperature

Potyrailo et al., *J. Opt.*, **2018**
Potyrailo et al., *ECS Transactions* **2019**
Potyrailo et al., *Faraday Transactions* **2020**



Biomimetics -
recreation of observed functionality
Room temperature

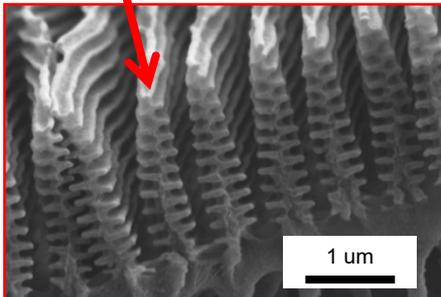
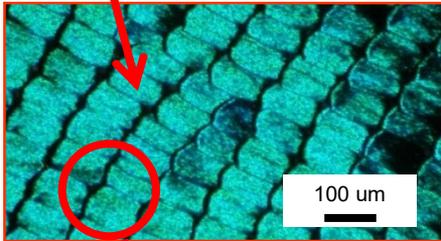
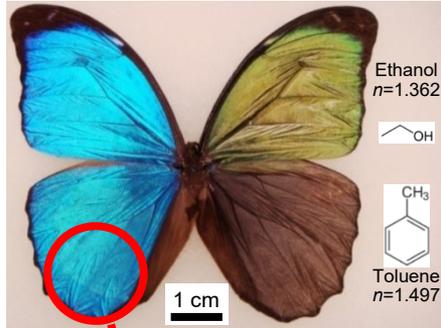
Potyrailo et al., *Nature Communications* **2015**



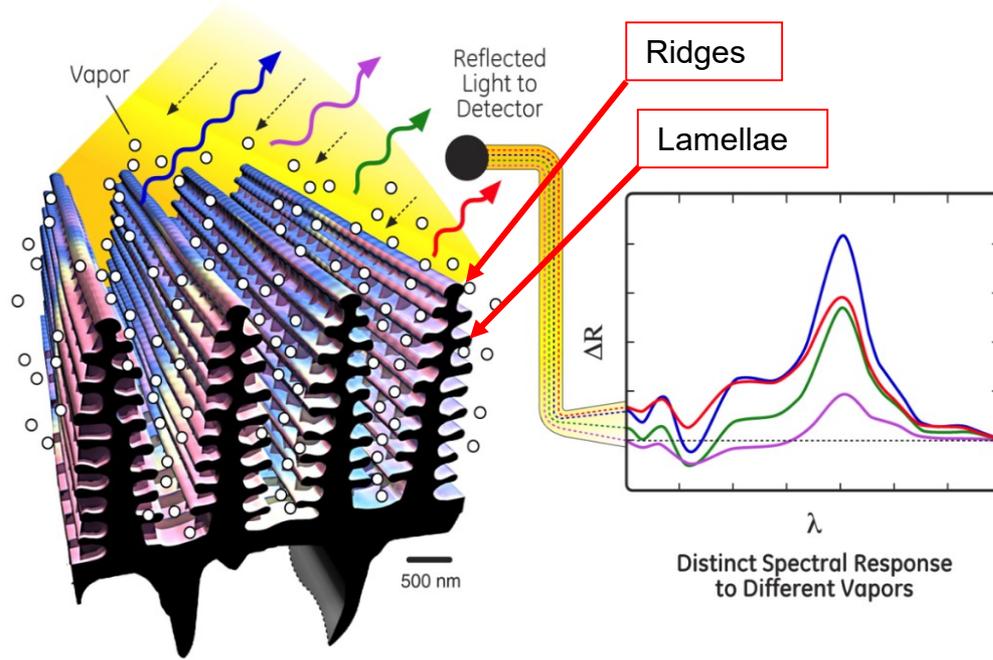
Biomimicry -
imitation of biological systems

Potyrailo et al., *Nature Photonics* **2007**
Potyrailo et al., *Proc. Natl. Acad. Sci. U.S.A.* **2013**

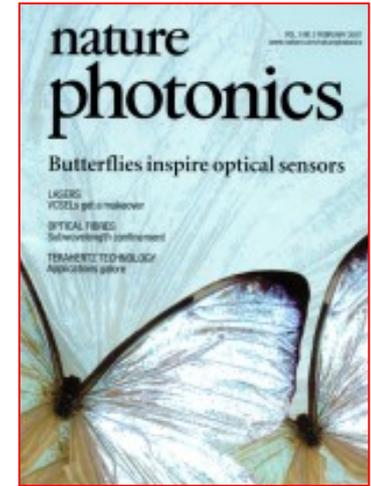
Structural color in nature: from understanding to functional applications



Operation principle of multivariable sensors utilizing natural *Morpho* butterfly scales

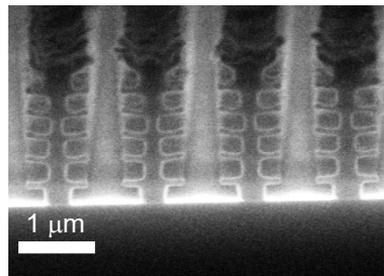


Potyrailo et al., *Nature Photonics*, 2007



Research curiosity brings a potential for useful performance

Single multivariable sensor outperforms classic QCM and MOS sensor arrays



nature COMMUNICATIONS

ARTICLE

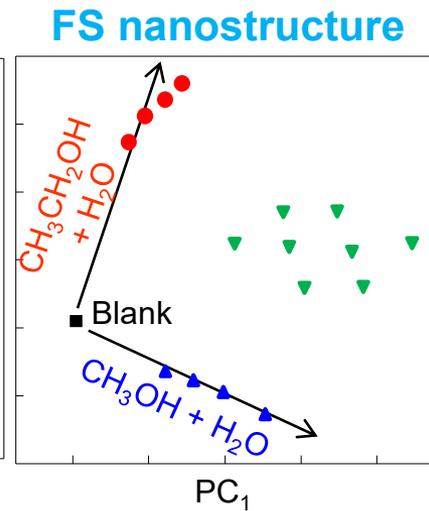
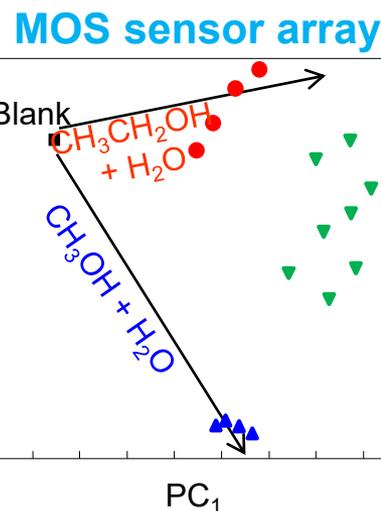
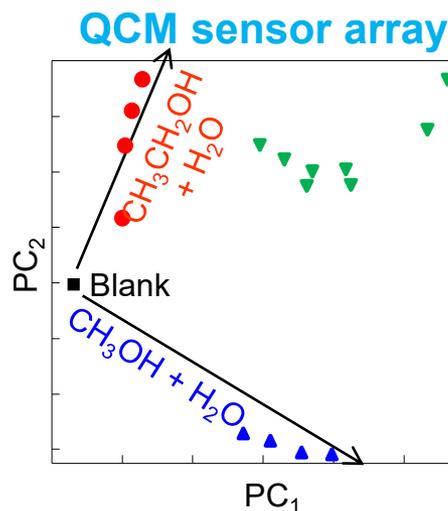
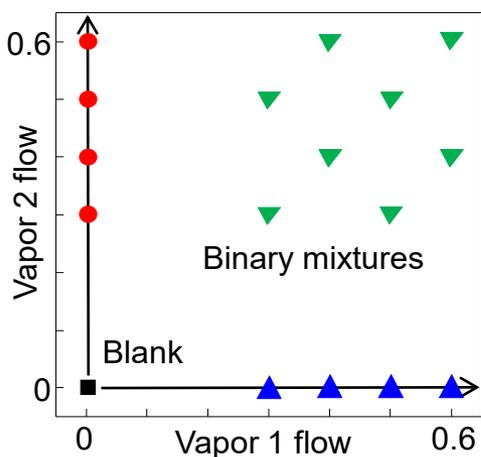
Received 22 Nov 2014 | Accepted 1 Jul 2015 | Published 1 Sep 2015

DOI: 10.1038/ncomms8959 OPEN

Towards outperforming conventional sensor arrays with fabricated individual photonic vapour sensors inspired by *Morpho* butterflies

Radislav A. Potyrailo¹, Ravi K. Bonam², John G. Hartley², Timothy A. Starkey³, Peter Vukusic³, Milana Vasudev^{4,5}, Timothy Bunning⁴, Rajesh R. Naik⁴, Zhexiong Tang¹, Manuel A. Palacios¹, Michael Larsen¹, Laurie A. Le Tarte¹, James C. Grande¹, Sheng Zhong¹ & Tao Deng^{1,6}

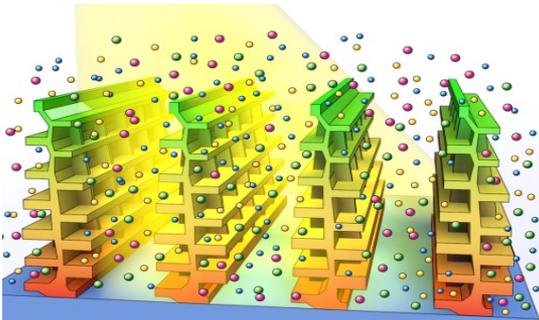
FS = nonafluorohexyl-trimethoxysilane
 QCM = quartz crystal microbalance
 MOS = metal oxide semiconductor



Ethanol, methanol vapors, their mixtures + water background

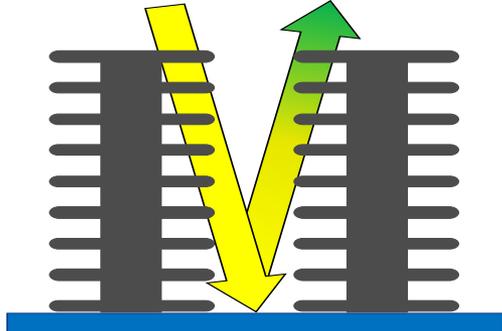
Advancing design rules of nanostructures for high temperature SOFC gas-sensing applications

Selectivity control for vapors at room temp.



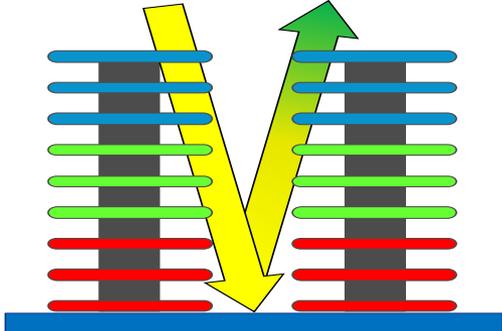
- Polymeric nanostructure
- Absorption and adsorption of vapors

Selectivity control for gases at high temp.

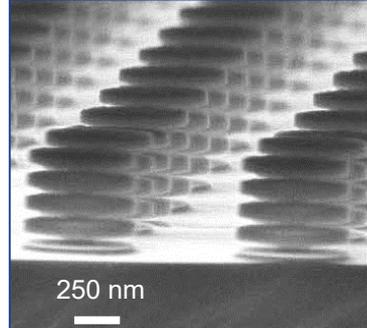
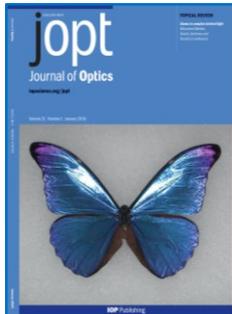


- Inorganic nanostructure
- Catalytic reactions of gases

Interference rejection control



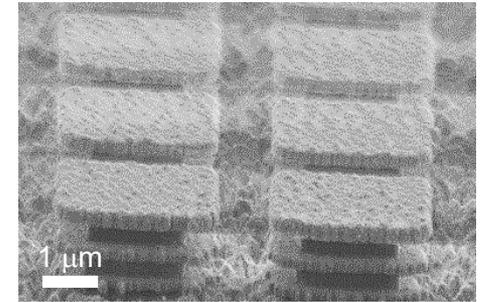
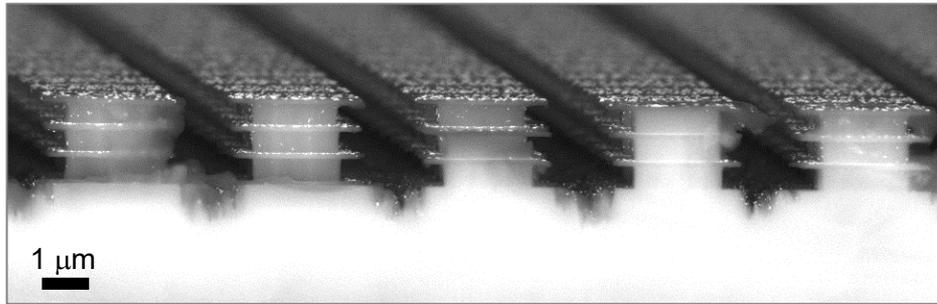
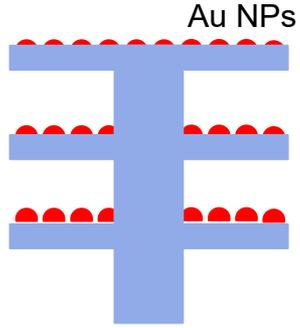
- Multi-material inorganic nanostructure
- Catalytic reactions of gases



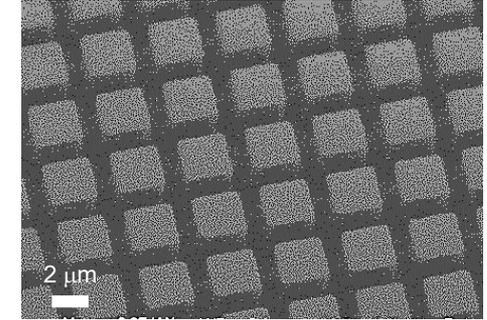
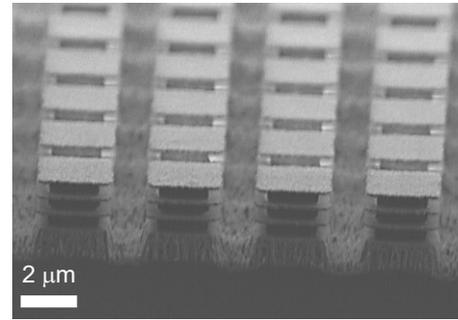
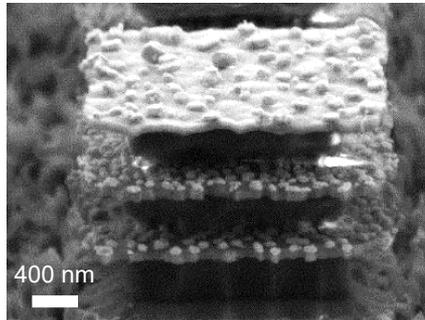
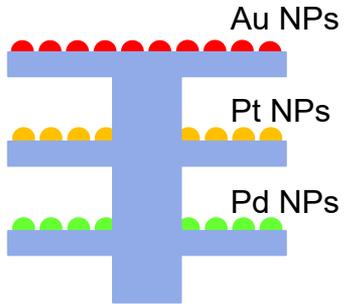
Potyrailo et al., *J. Opt.*, 2018
Potyrailo et al., *ECS Transactions* 2019
Potyrailo et al., *Faraday Transactions* 2020

Nanostructures for gas sensing at high temperature

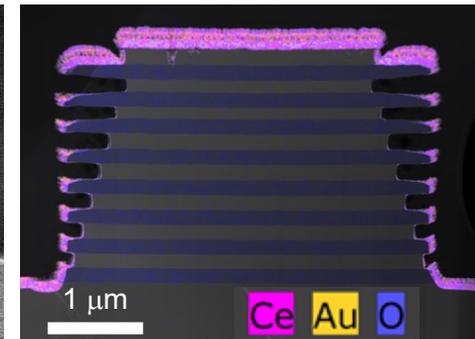
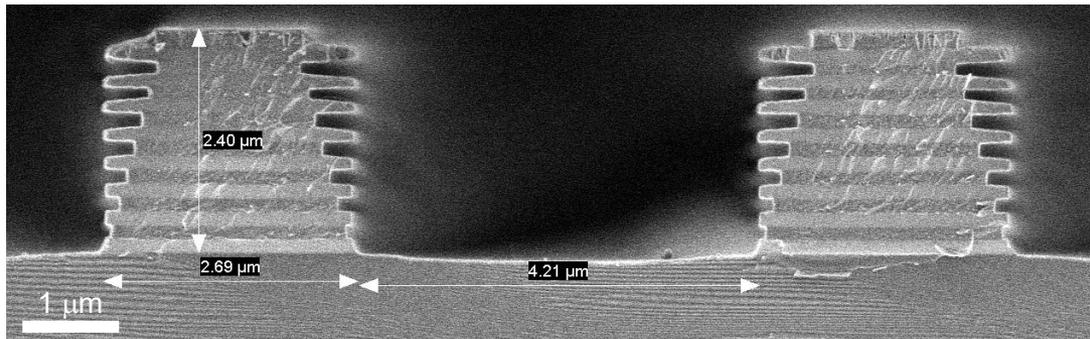
Nanostructure with Au nanoparticles before and after capping with CeO₂ layer



Nanostructure with Pd, Pt, and Au nanoparticles before capping with CeO₂ layer



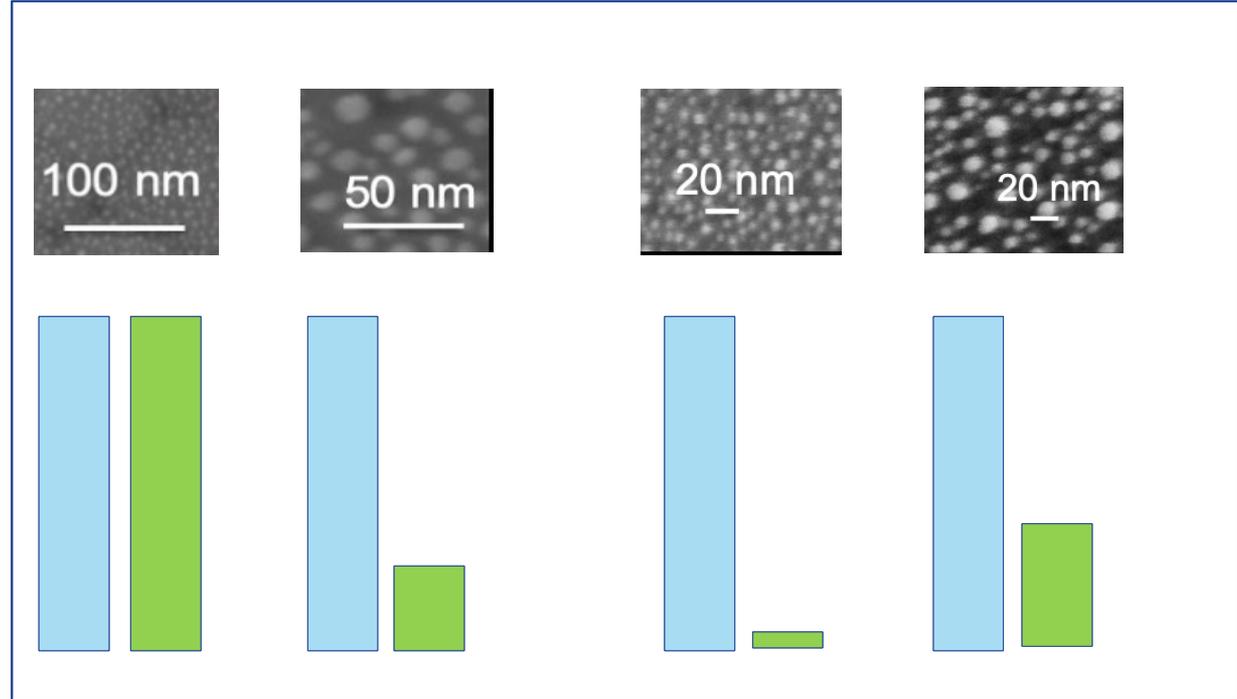
Nanostructure with Au nanoparticles after capping with CeO₂ layer



Size and metal type effects on diversity of sensitivity to H₂ and CO



Relative sensitivity H₂ to CO



Au NPs
~2 - 5 nm

Au NPs
~6 - 10 nm

Pt NPs
~2 - 5 nm

Pt NPs
~6 - 10 nm

Experimental data:
Operation temperature 300 °C
Sensitivity ratios over ~520 – 860 nm
Scanning electron microscope images are representative examples

Data: M. Carpenter, SUNY Poly

Machine learning support:

Data analytics =

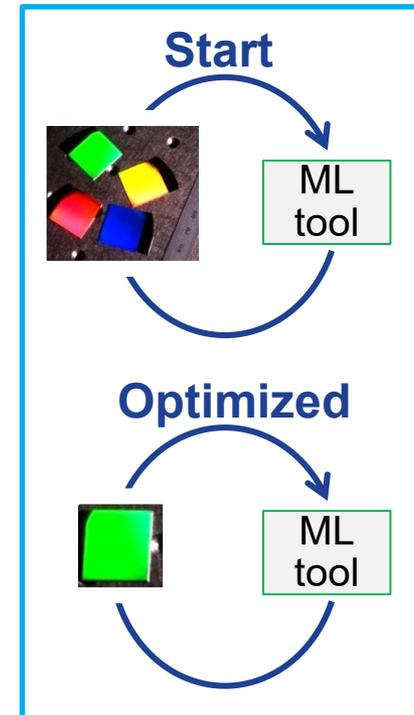
a.k.a. chemometrics, multivariate statistics, machine learning (ML)

Examples of our ML tools

- Support Vector Machines (SVM)
- Principal component analysis (PCA)
- Hierarchical cluster analysis (HCA)
- Discriminant Analysis (DA)
- Artificial Neural Network (ANN)
- Independent Component Analysis (ICA)
- Partial least squares (PLS) regression
- Principal Component Regression (PCR)
- **New tools for boosting sensor stability**

R. A. Potyrailo, Multivariable sensors for ubiquitous monitoring of gases in the era of Internet of Things and Industrial Internet, *Chem. Rev.* **2016**, *116*, 11877–11923

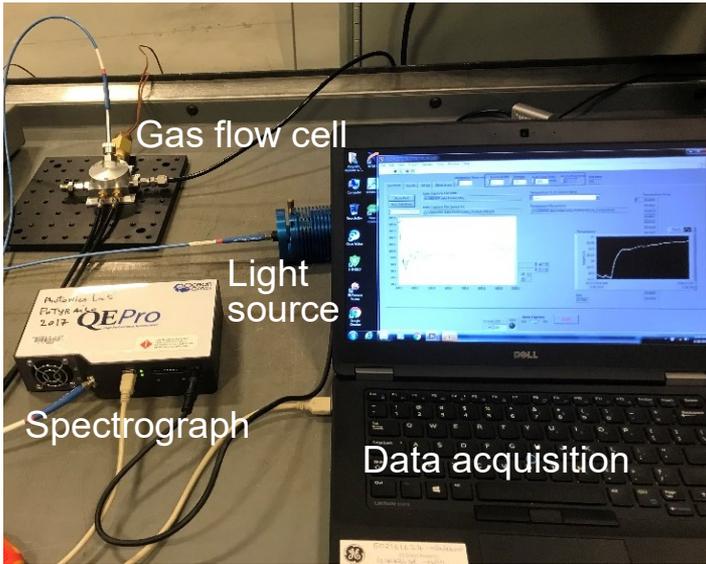
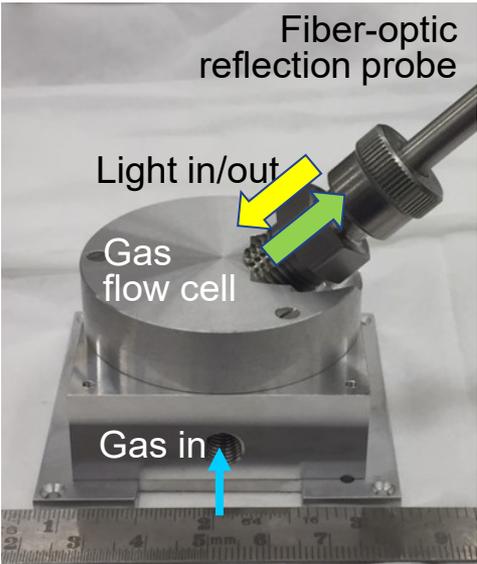
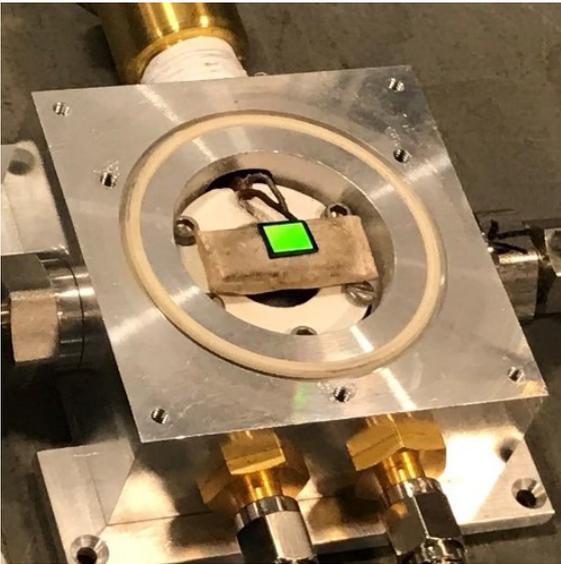
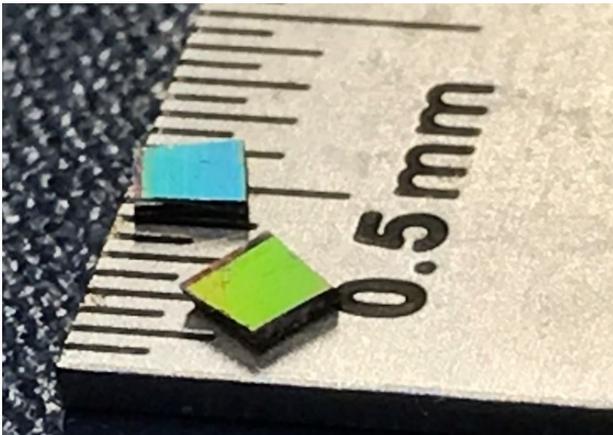
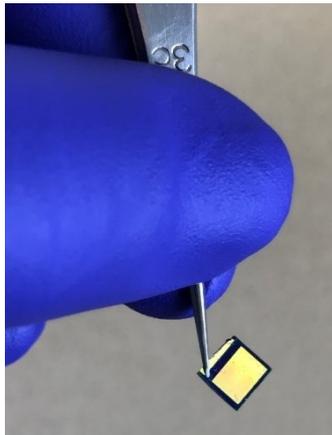
ML
for sensor system optimization



Potyrailo et al., *Faraday Transactions* **2020**

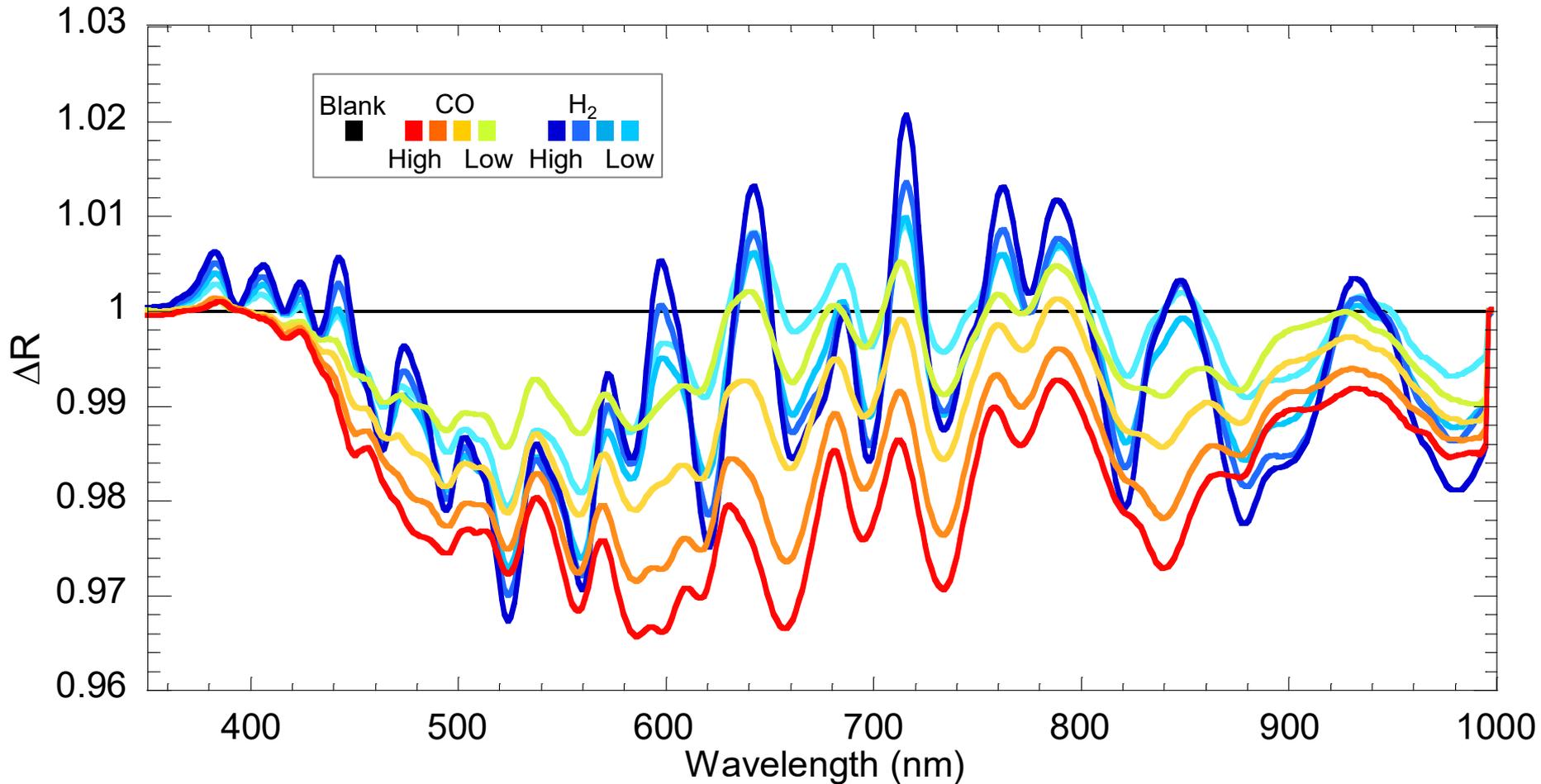
Increasing role of data analytics in high performance sensing

Fabricated bio-inspired photonic 3-D nanostructures and their gas testing



Spectral response to H₂ and CO

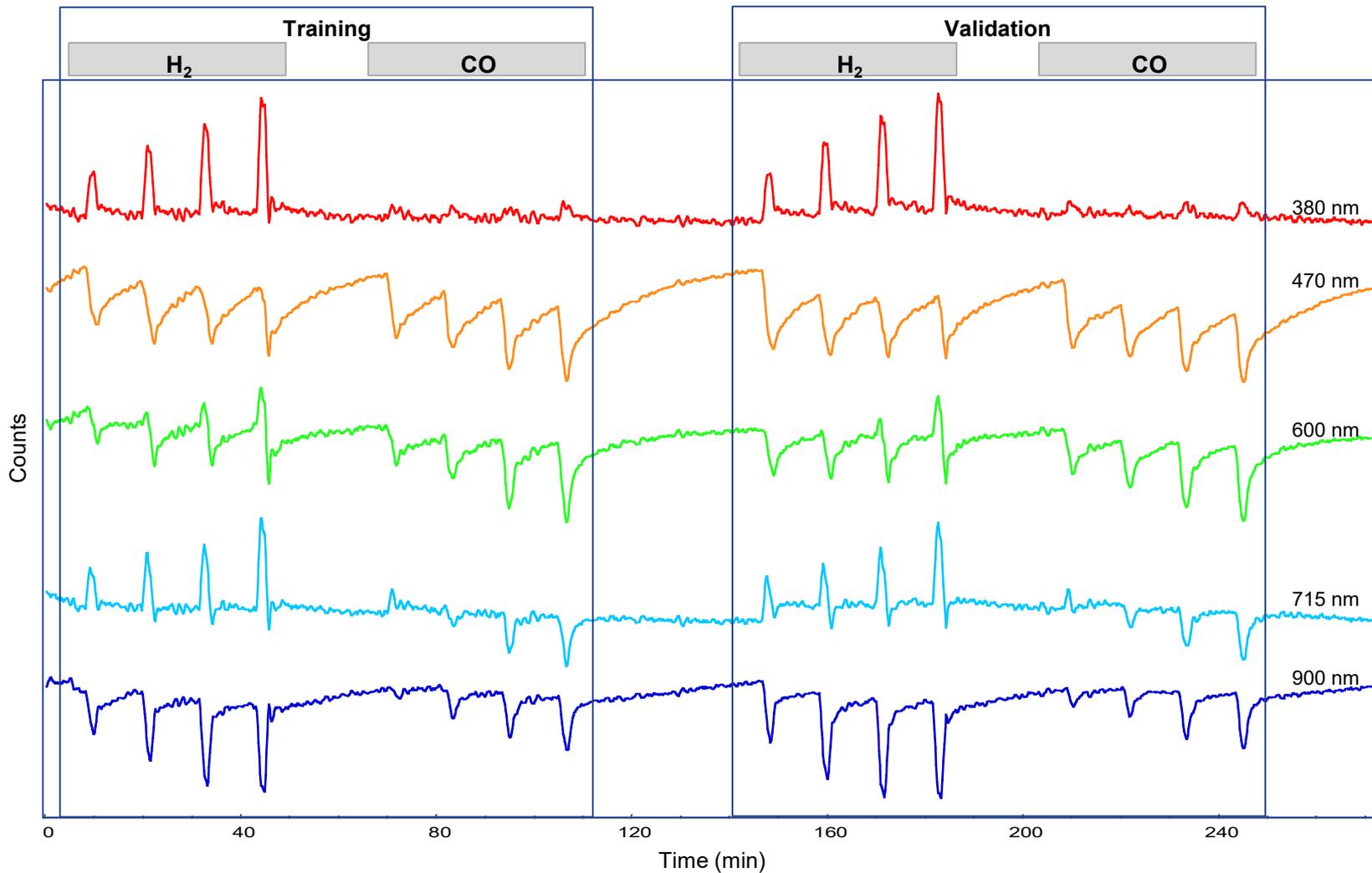
H₂ and CO conc:
10, 20, 30, 40 %



Spectral diversity of responses at different wavelengths
allows discrimination of H₂ and CO gases

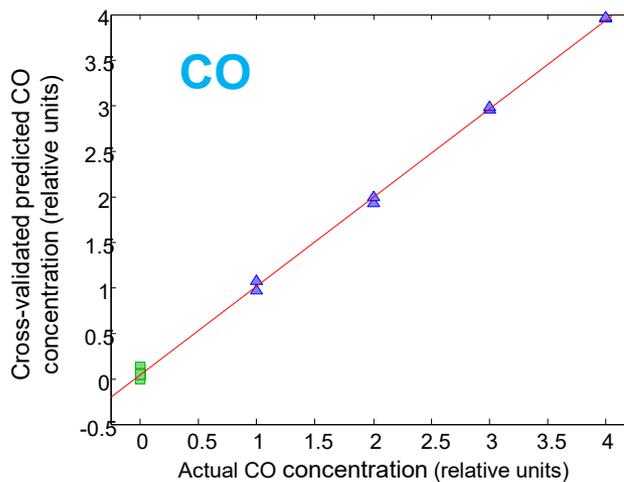
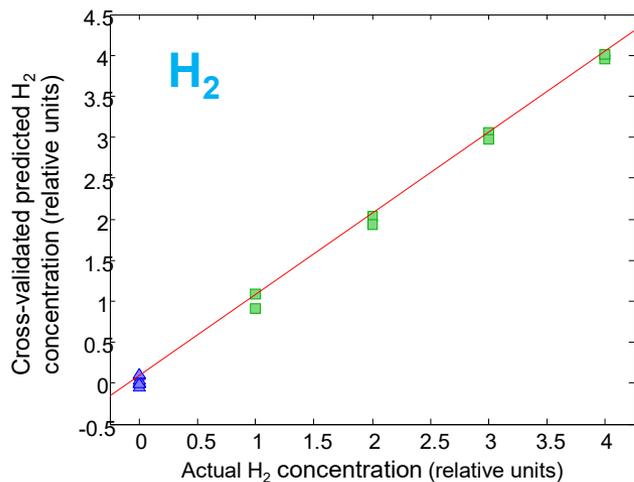
Dynamic response to H₂ and CO

H₂ and CO conc:
10, 20, 30, 40 %

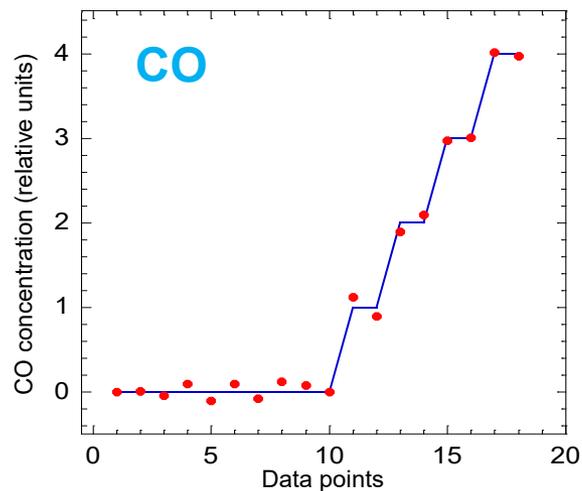
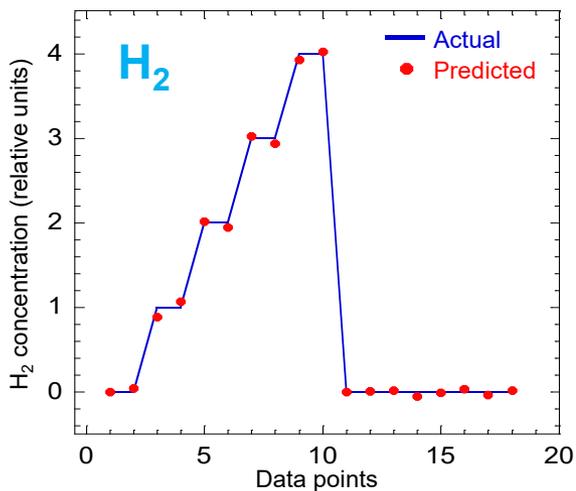


Discrimination of H₂ and CO based on diversity of responses at different wavelengths:
(1) directions of the response, (2) relative response intensities

Cross validated prediction of H₂ and CO



Coded 1 – 4:
H₂ and CO conc:
10, 20, 30, 40 %

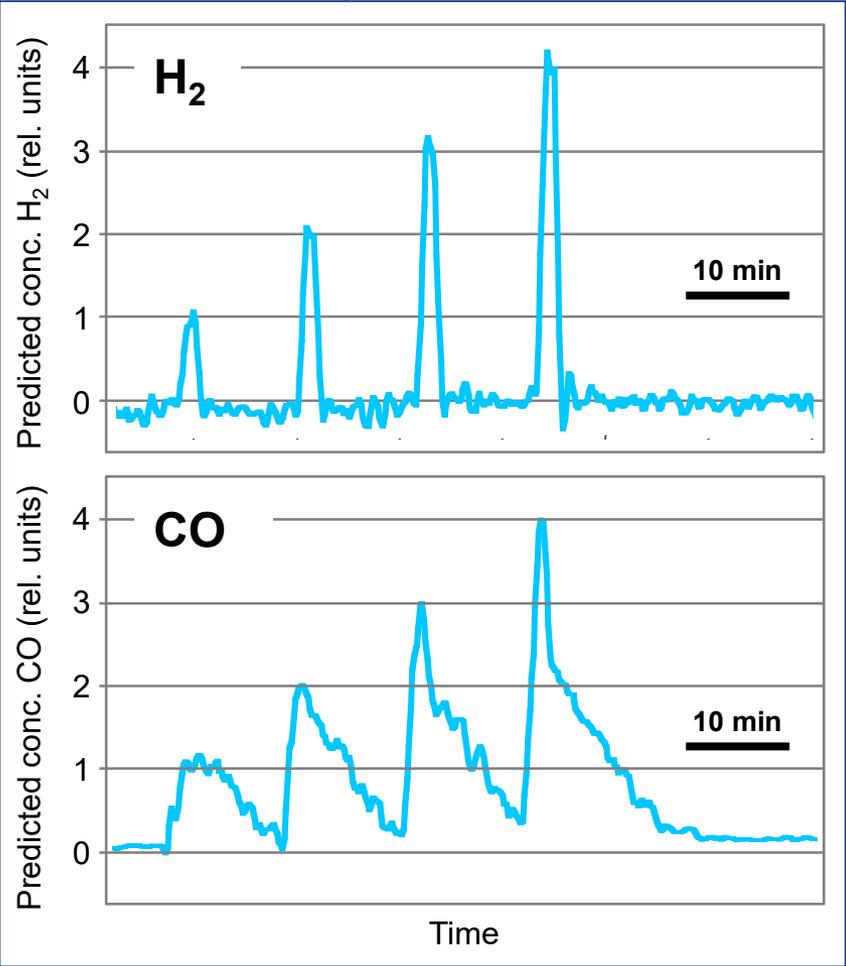


Initially tested sensor ability to predict concentrations of H₂ and CO gases

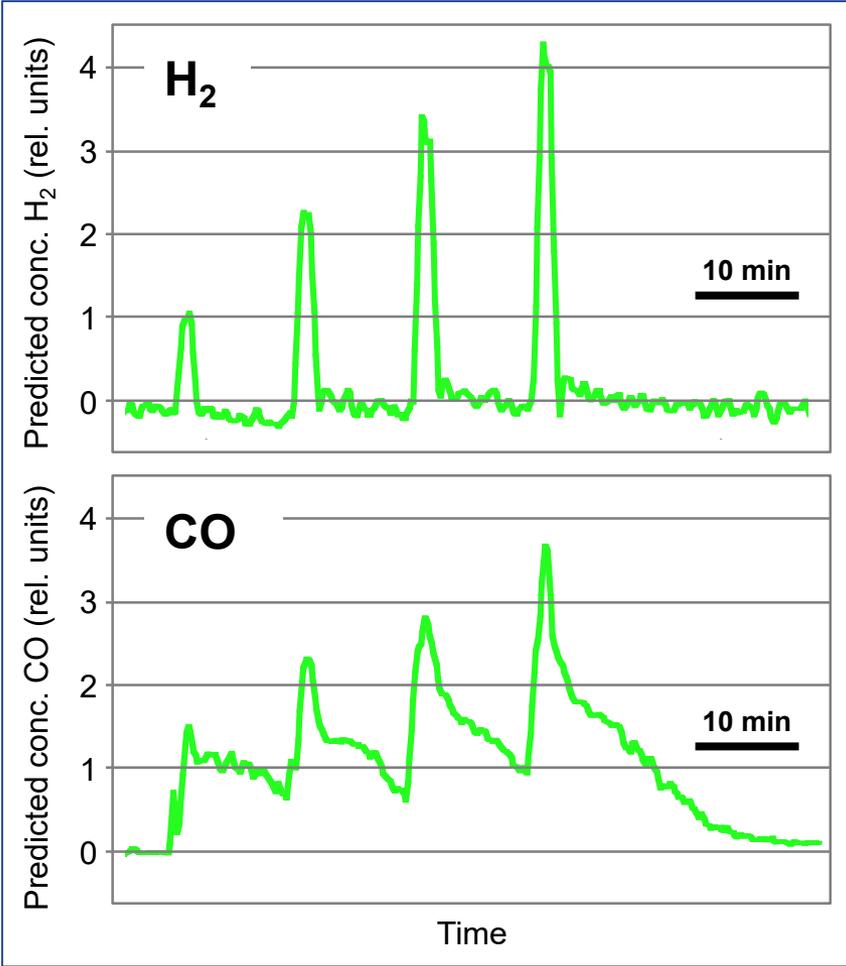
Dynamic prediction of H₂ and CO

Coded 1 - 4:
H₂ and CO conc:
10, 20, 30, 40 %

Training

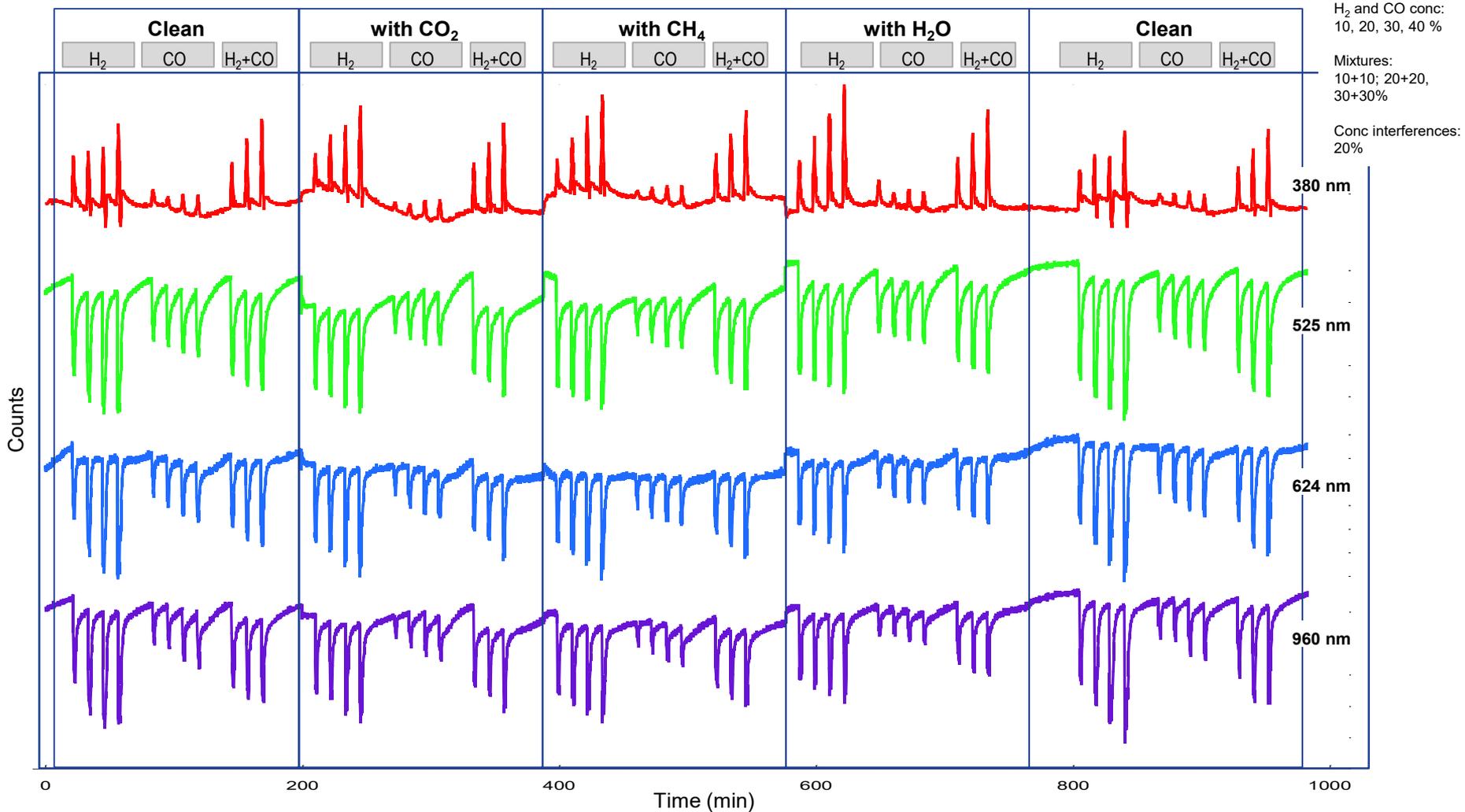


Validation



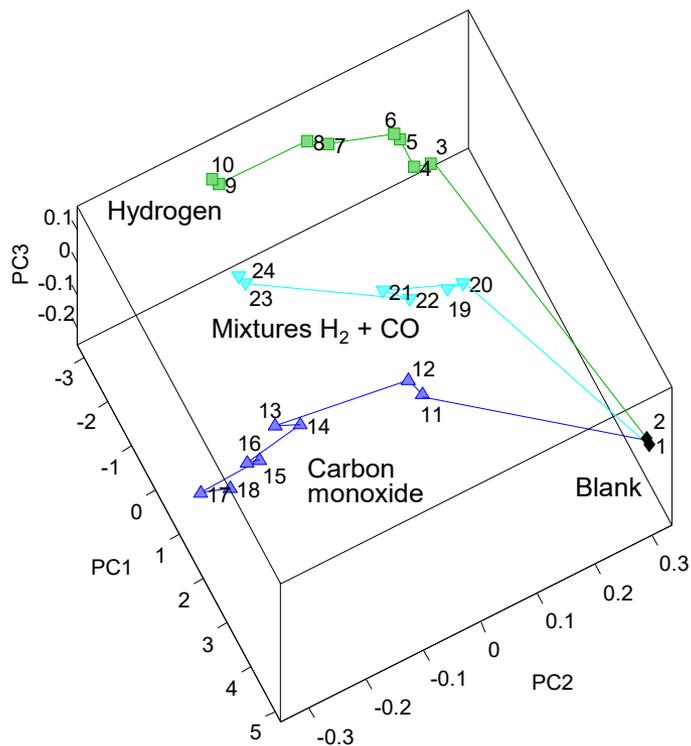
Initially tested sensor ability to predict concentrations of H₂ and CO gases

Discrimination of H₂ and CO and their mixtures and with interferences CO₂, CH₄, and H₂O

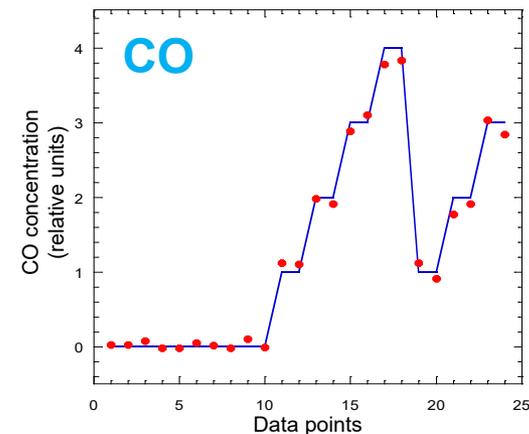
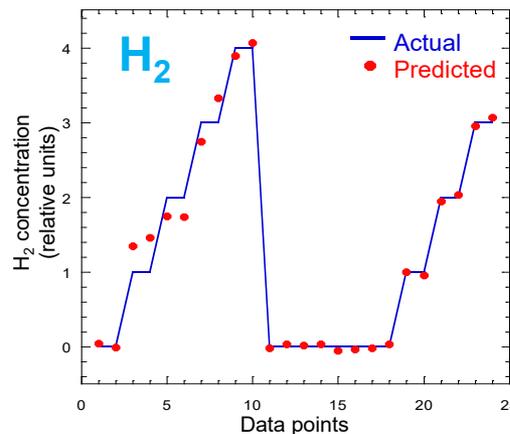
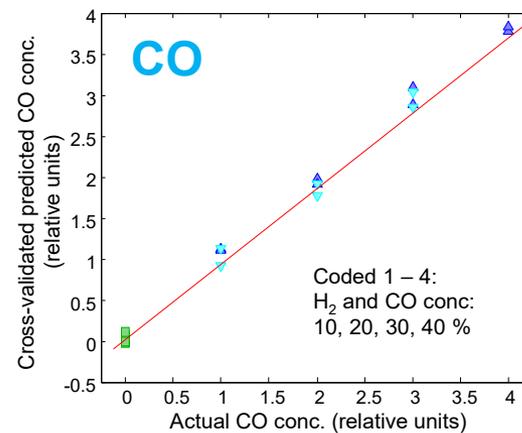
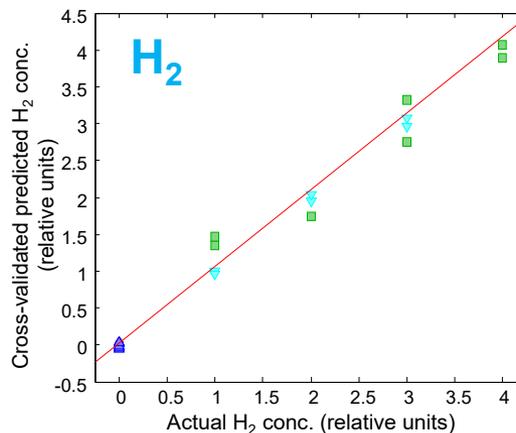


Rejection of interferences is a must in sensor performance

Quantitation of H₂ and CO in mixtures

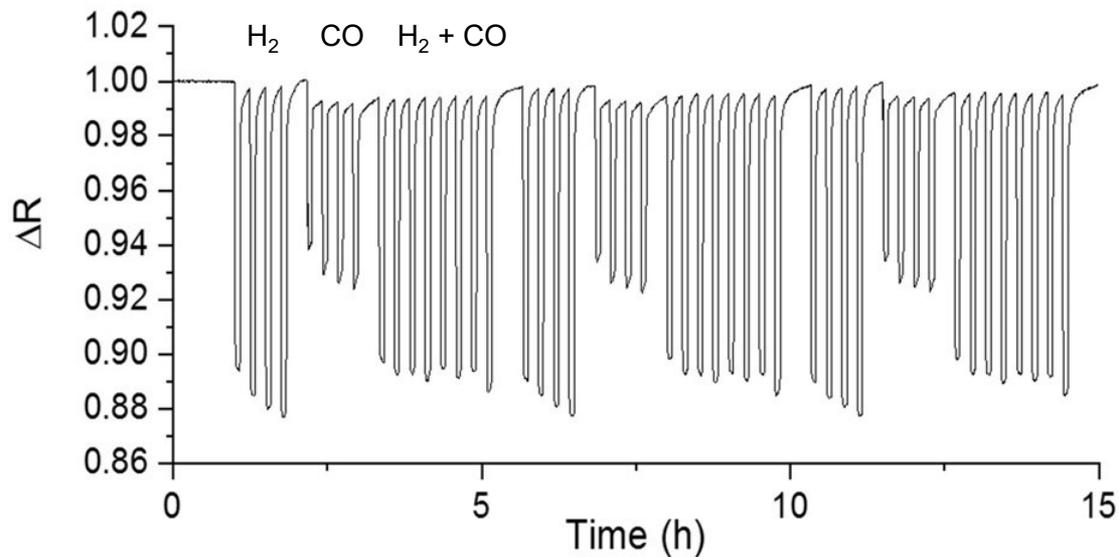


H₂ and CO conc: 10, 20, 30, 40 %
 Mixtures: 10+10; 20+20, 30+30%
 Conc interferences: 20%

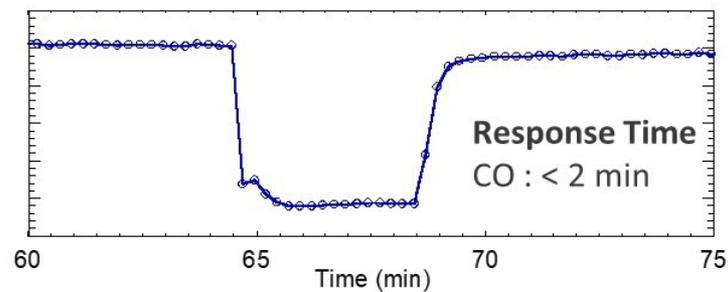
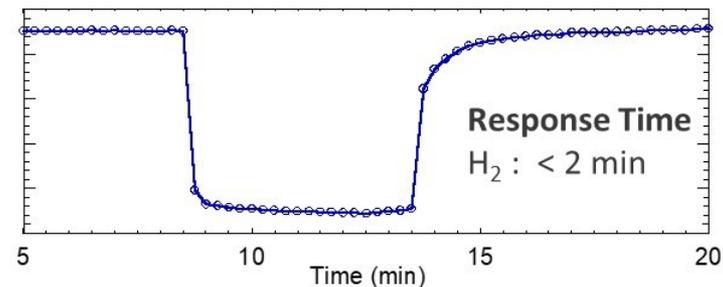


Initially demonstrated discrimination of individual gases and their mixtures
 = a must in sensor performance

Example of response stability test



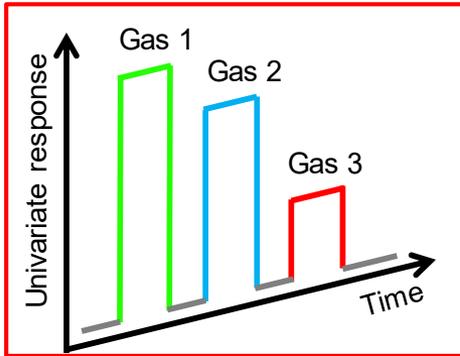
Stability: Baseline drift: <1% full scale over 15 hours



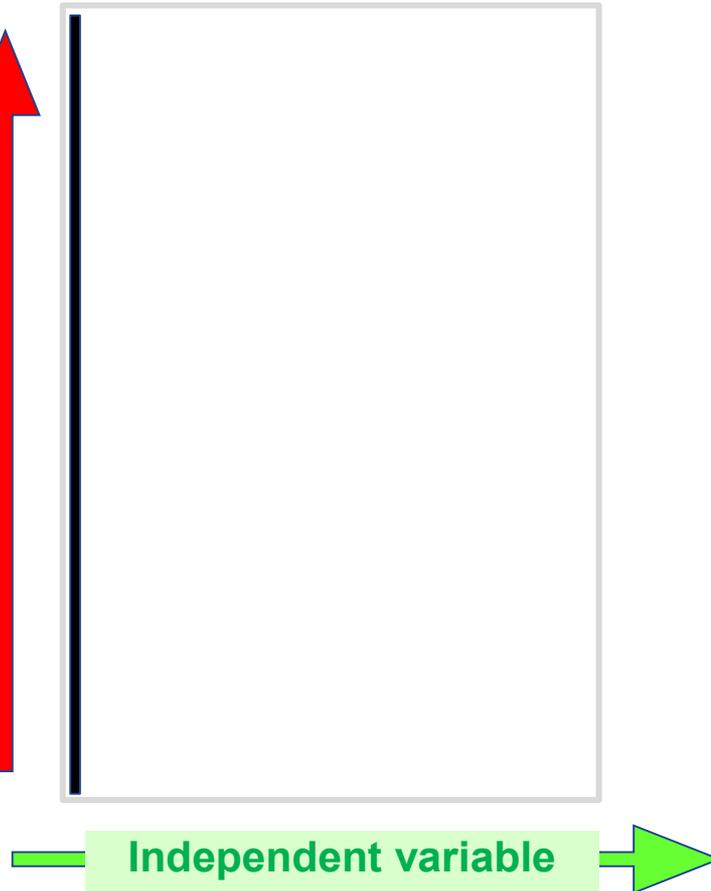
Data: M. Carpenter, SUNY Poly

GE Vision: Boosting sensor response dimensionality

Conventional
single-output sensor

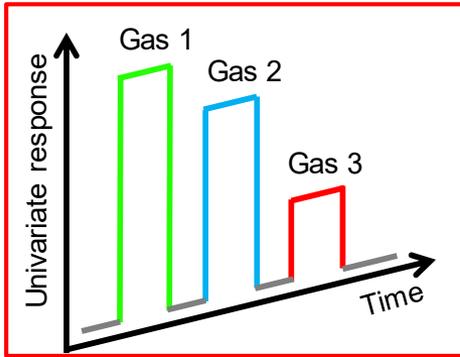


Single-output response



GE Vision: Boosting sensor response dimensionality

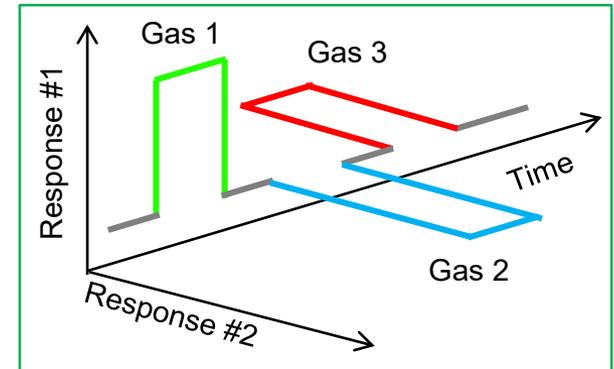
Conventional single-output sensor



Single-output response



Developed multi-output sensor

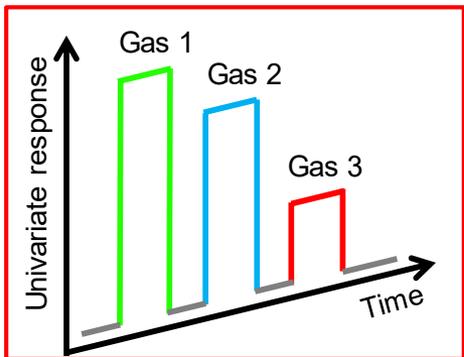


Independent variable

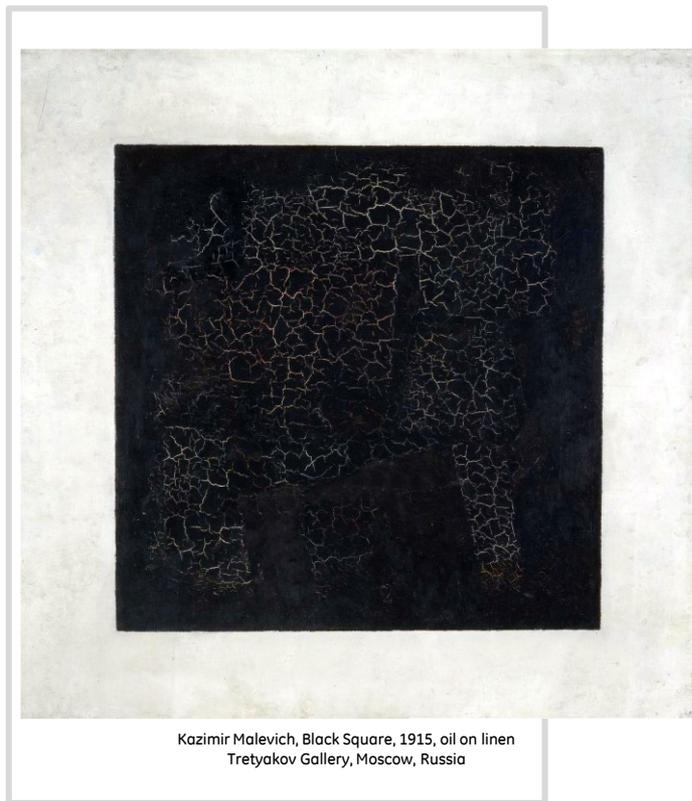
Technology accumulation in diverse areas provide previously untapped opportunities to build new generation of sensors with complementary capabilities to traditional analytical instruments

Importance of correct independent variables in design of multivariable sensors

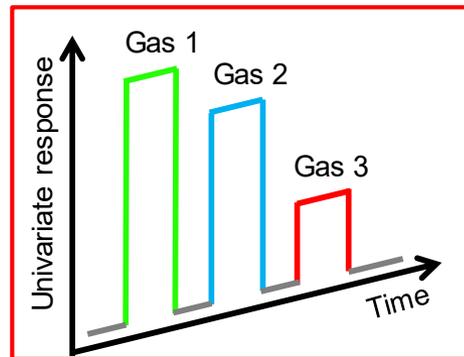
Conventional single-output sensor



Single-output response



Independent variable



Developed design rules of nanostructures for high temperature gas-sensing applications

1) Gas-sensitivity:

- Catalytic reactivity of plasmonic nanoparticles incorporated in metal oxide matrix
- Gas responses driven by changes in the refractive index and extinction coefficient

2) Functions of 3D nanostructure:

- High surface area for interactions with ambient gases
- Spectral discrimination of catalytic reactions in different regions of 3D nanostructure

3) Gas-selectivity:

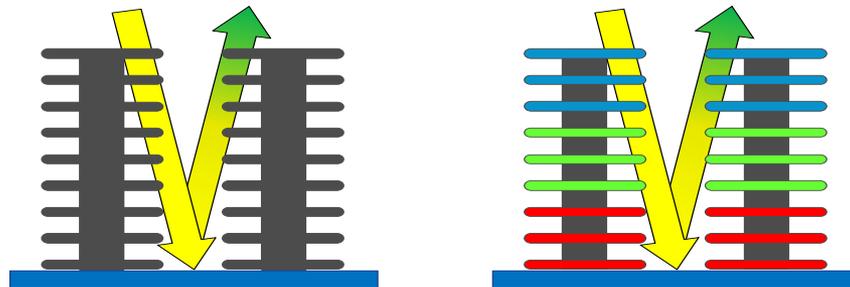
- Diversity of catalytic activity of metal nanoparticles
- Origins of catalytic activity diversity of metal nanoparticles: size, metal type, metal oxide type
- Spatial distribution of catalytically diverse nanoparticles for wavelength-dependent gas response

4) Diverse catalytic activity of metal nanoparticles on 3D nanostructure:

- Nanoparticles of the same noble metal but of different sizes
- Nanoparticles of the different noble metals

5) Rejection of gas interferences:

- Spectral discrimination of response of 3D nanostructure to diverse gases



Acknowledgments



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