

Strategies to enhance the selectivity in high temperature gas sensing

Yu Lei

Department of Chemical and Biomolecular Engineering

University of Connecticut

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Outline

Background

Research Objective

**Strategies to Improve the High
Temperature Gas Sensing Selectivity**

Conclusions and Outlook

Background

Importance of Harsh Environment Gas Sensors

Environment & Energy Concerns

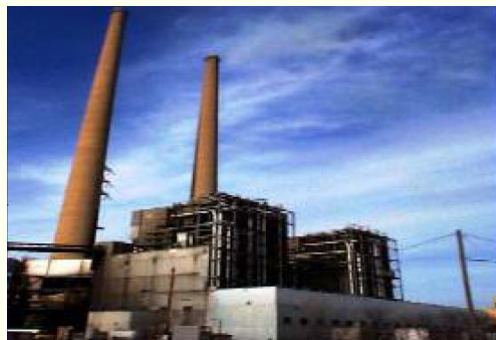
- Better control of **combustion**
- Reduction of **emissions** (CO, NO_x, SO_x, HCs) → less environmental problems
- Improvement of **energy efficiency** → more energy savings

❖ Contribution from HTGS and Controls

Value Derived for an Existing Coal Fired Power Plant

FUEL COSTS

\$39 Million/Year



→ Gaseous Emissions

POWER

→ 3,285,000,000 kw-hr/yr
@75% Capacity Factor

→ Solid Wastes

Total Fuel + O&M Budget \$45 Million - Avg. 500 MW Unit (Analysis for 2000)

➤ 1% improvement in EFFICIENCY

• \$390,000 savings in fuel per year

600

million for entire installed fossil capacity

➤ 1% REDUCTION
in greenhouse gases
and solid wastes

DOE prediction:

Energy savings per year

0.25 quadrillion BTU



- 11 million tons of coal
- 250 billion cubic feet of natural gas
- 43 million barrels of crude oil



Background

Sensors for Harsh Environments



Solid Oxide Fuel Cells

- 650 – 1000 °C
- Atmospheric pressure

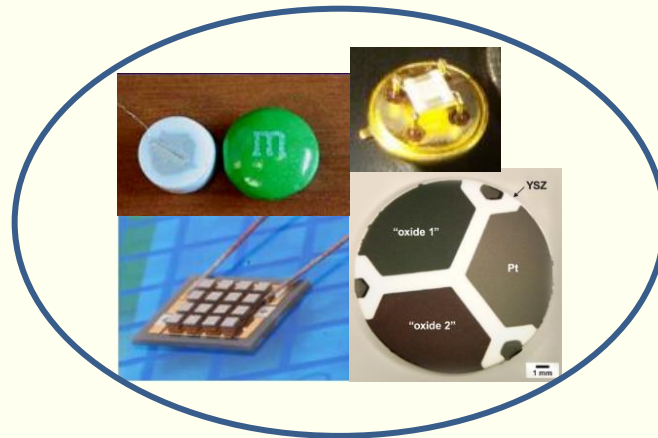


Advanced Combustion Turbines

- Up to 1300 °C combustion temperatures
- Pressure ratios of 30:1

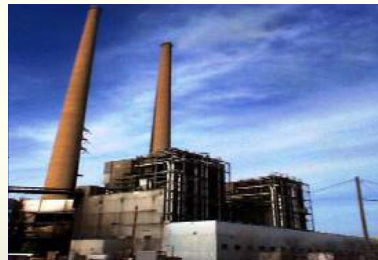
Automotive Engine

- up to 1000 °C
- Compression ratio ~10:1



Ultra Supercritical Boilers

- Up to 760 °C temperature
- Up to 5000 PSI pressure



Gasifiers

- Up to 1600 °C, and 1000 PSI (slagging gasifiers)
- Erosive, corrosive, highly reducing environment



Background

Sensors for Harsh Environments

❖ Applications Harsh Industrial Environment

- Heat treating
- Glass & Ceramic
- Automotive
- Aerospace
- Pulp and paper
- Utility and power
- Metal processing and casting
- Chemical and petrochemical
- Food processing

Table 1. Example of exhaust conditions for two- and four-stroke, diesel and lean-four-stroke engines

| Exhaust components and conditions ^a | Diesel engine | Four-stroke spark ignited-engine | Four-stroke lean-burn spark ignited-engine | Two-stroke spark ignited-engine |
|--|---------------------|----------------------------------|--|---------------------------------|
| NO _x / ppm | 350–1000 | 100–4000 | ≈ 1200 | 100–200 |
| HC / ppm C | 50–330 | 500–5000 | ≈ 1300 | 20,000–30,000 |
| CO | 300–1200 ppm | 0.1–6% | ≈ 1300 ppm | 1–3% |
| O ₂ | 10–15% | 0.2–2% | 4–12% | 0.2–2% |
| H ₂ O | 1.4–7% | 10–12% | 12% | 10–12% |
| CO ₂ | 7% | 10–13.5% | 11% | 10–13% |
| SO _x / ppm | 10–100 | 15–60 | 20 | ≈ 20 |
| Temperatures / °C | r.t.–650 (r.t.–420) | r.t.–1100 ^b | r.t.–850 | r.t.–1000 |
| λ (A/F) ^c | ≈ 1.8 (26) | ≈ 1 (14.7) | ≈ 1.16 (17) | ≈ 1 (14.7) |



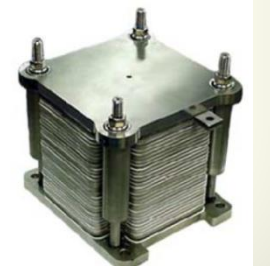
~1000 °C

Automobile Engine



Up to 1300 °C

Advanced Combustion Turbines



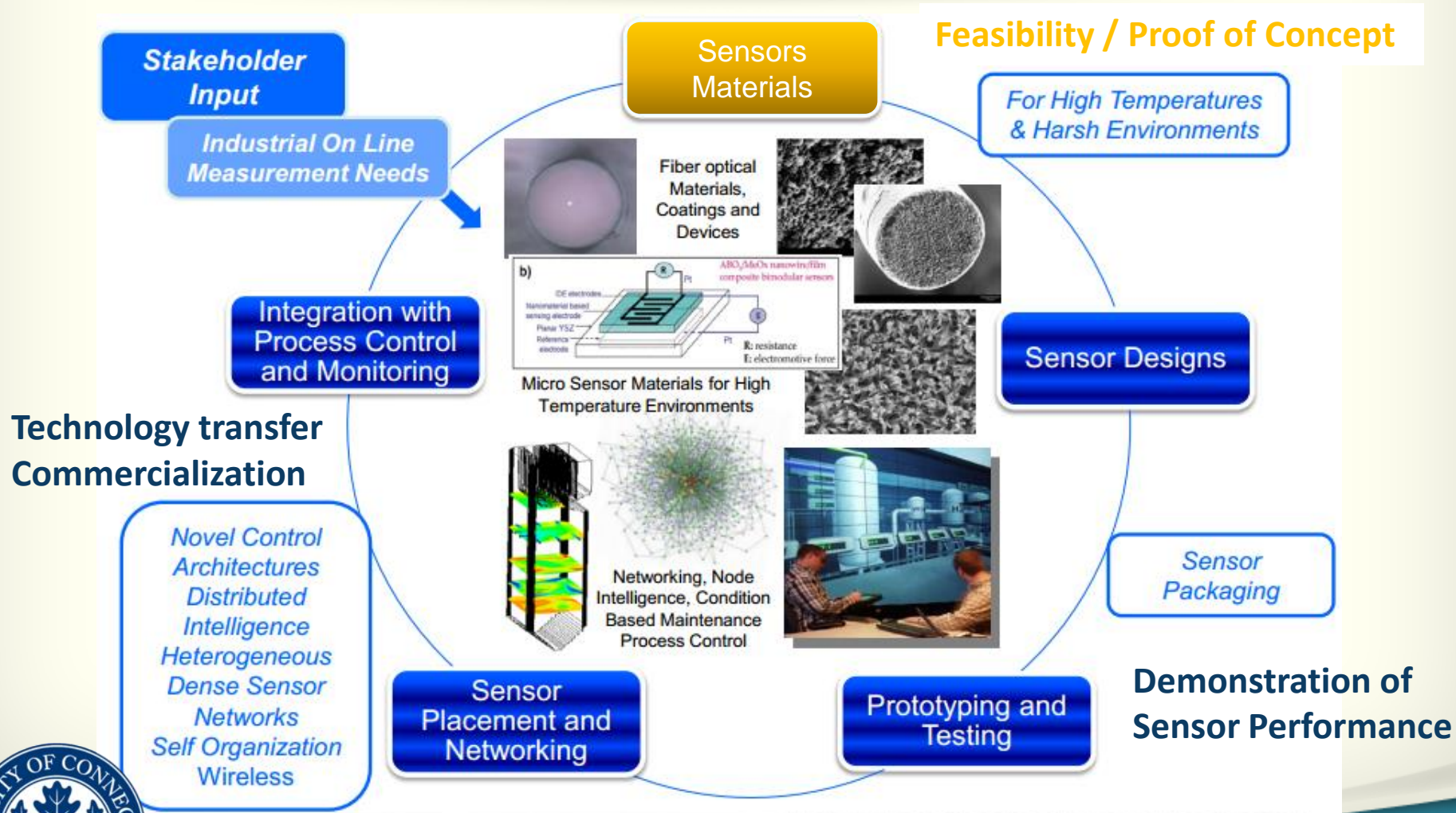
650 - 1000 °C

Solid Oxide Fuel Cell



Background

Research in Sensors and Controls



Background

Current Solid-State Sensing Technologies

Potentiometric

- Commercialized O₂ sensor (Zirconia)
- CO₂, SO_x (K₂CO₃, Ag₂SO₄) at mild T

Amperometric

- Diffusion limited current (linear)
- O₂, H₂, CO, NO_x, SO₂, H₂S

Resistive

- **Bulk conduction-based**
O₂ at 700 – 1100 °C
- **Surface conduction-based**
CO, HCs, H₂, NO_x in 200 – 500 °C
- **Metal/oxide junction based**
MS/MIS, MOSFET, Schottky diodes

Mixed Potential

- Thermodynamic non-equilibrium
- More than one electrochemical reaction occurs
- The oxidation and reduction reactions are for different species
- CO, NO_x, HCs in 500 – 600 °C

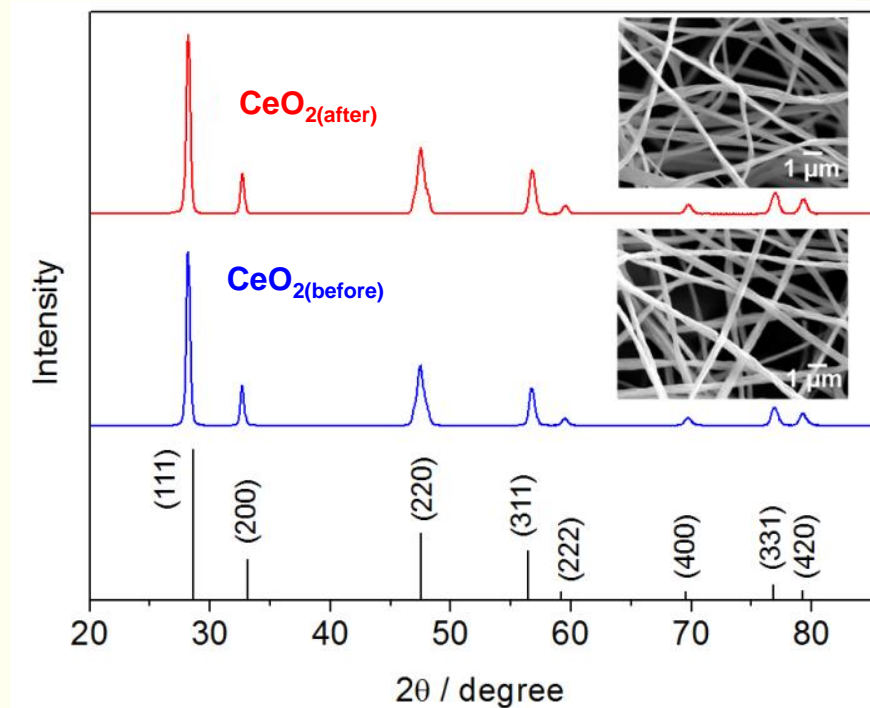
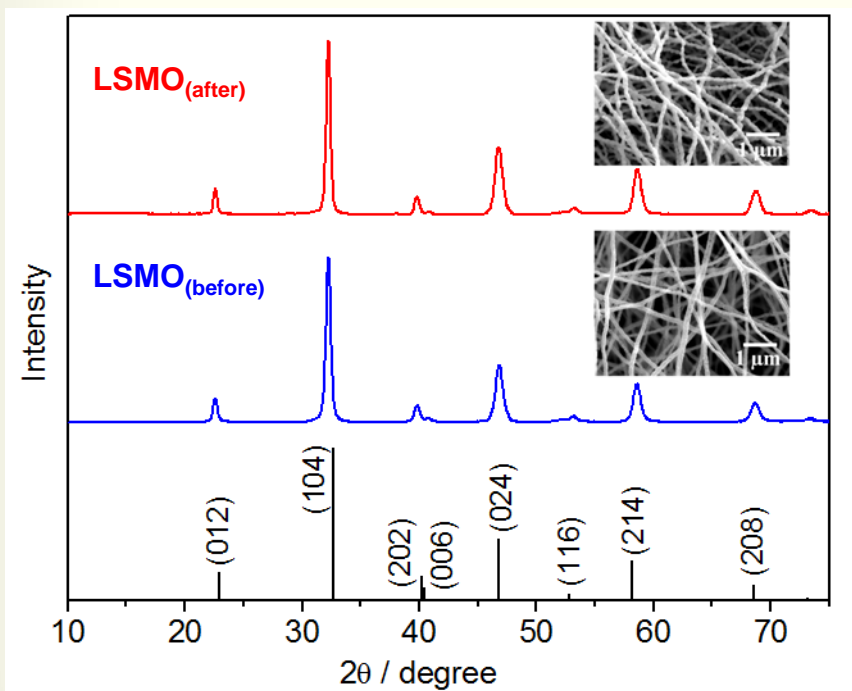
Impedancemetric

- AC measurements at a specified F
- Solid-electrolyte or resistor configured
- H₂O, CO, HCs, NO_x
- Measure total NO_x concentration
- Accurate detection on single ppm level



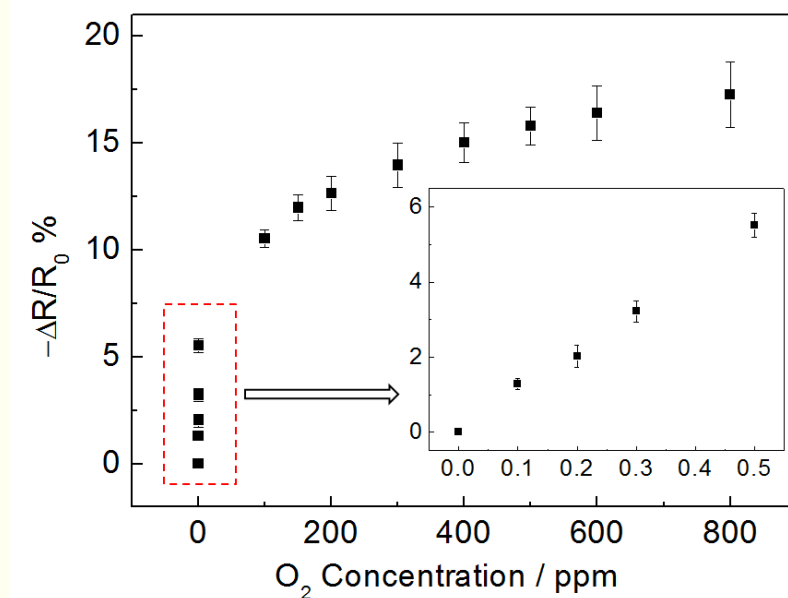
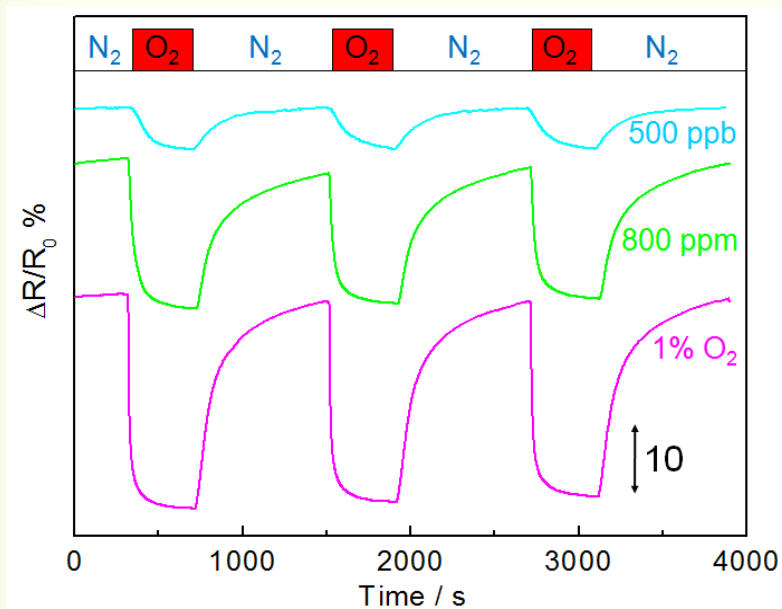
Easy to achieve highly stable and highly sensitive high temperature gas sensor

➤ Multiple cycles of heating/cooling treatment



Easy to achieve highly stable and highly sensitive high temperature gas sensor

LSMO NFs for O₂ Sensing at 800 °C

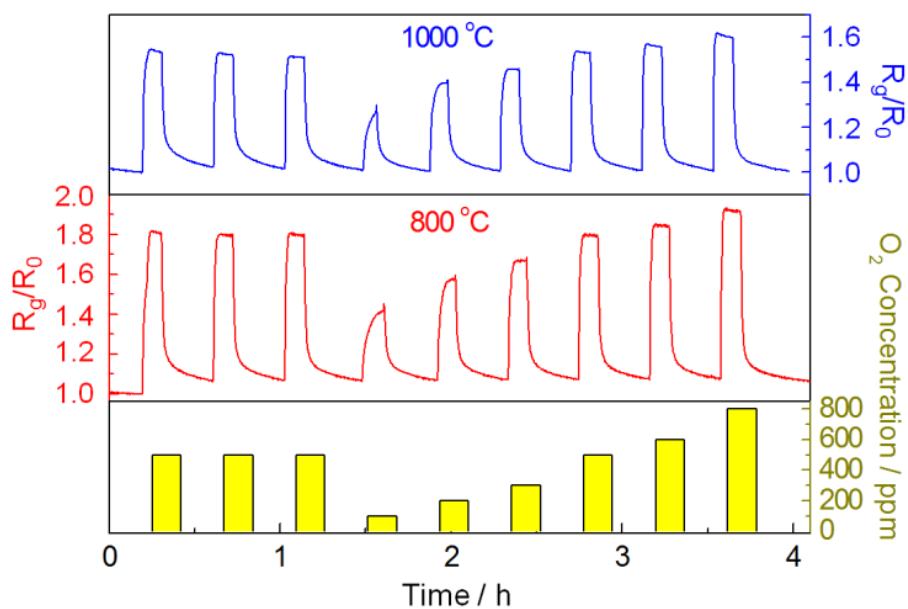


LSMO (p-type): R ↓ upon exposure to O₂

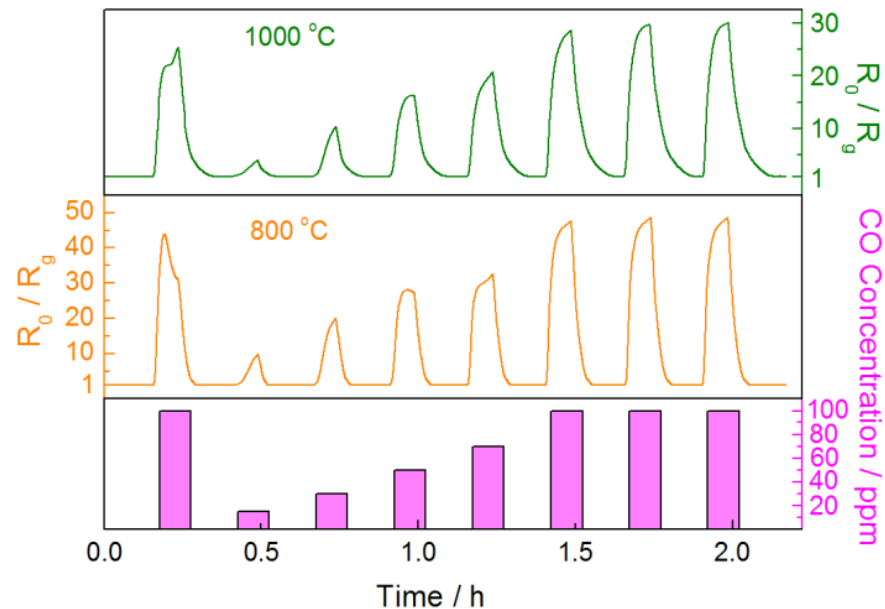


CeO₂ NFs for O₂ and CO Sensing at 800 and 1000 °C

Oxygen sensing



CO sensing



CeO₂ (n-type):

R ↑ upon exposure to O₂

R ↓ upon exposure to CO

➤ Sensitivity:

- In O₂ (oxidizing gas): R_g/R_0

- In CO (reducing gas): R_0/R_g

R_0 - initial resistance in N₂

R_g - measured real-time resistance

✓ Fast response

✓ Good recovery

✓ Good reproducibility

✓ High sensitivity



Research Objective

- **Selectivity** is the most pressing issue in high temperature gas sensing field
- Urgent need to develop gas sensors to achieve highly selective detection of gas species and concentration in high-temperature mixed gases environment

Objective

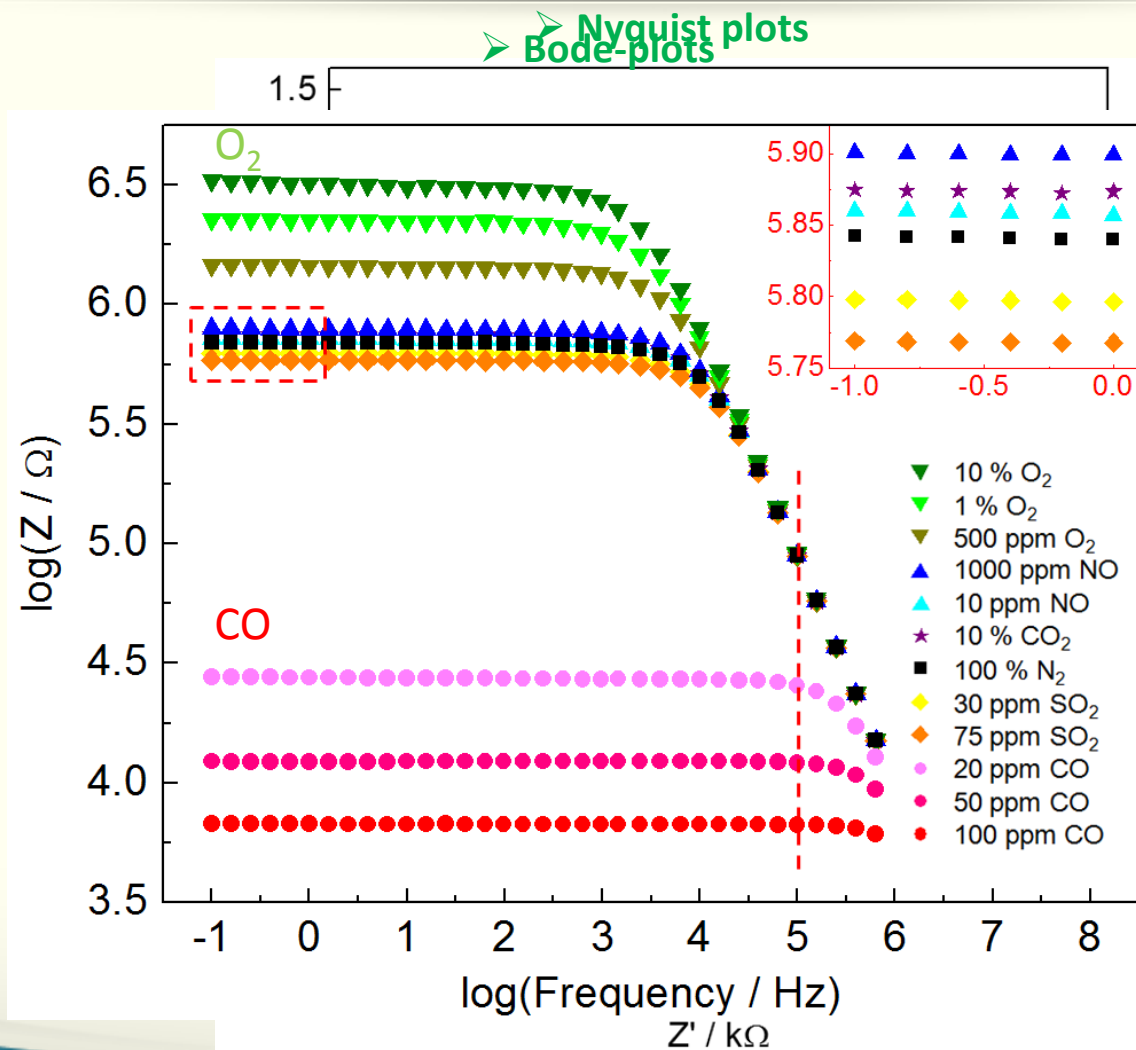
To address selectivity issue in high temperature gas sensing, following strategies are employed:

- Employing unique sensing technology for enhanced selectivity
- Engineering sensing materials for enhanced selectivity
- Developing machine learning algorithms to identify the gas concentration and species



I. Employing unique sensing technology for enhanced selectivity

Pt-CeO₂ NFs Impedancemetric Based Sensor



- Initial potential: 0 V
- Amplitude: 0.5 V



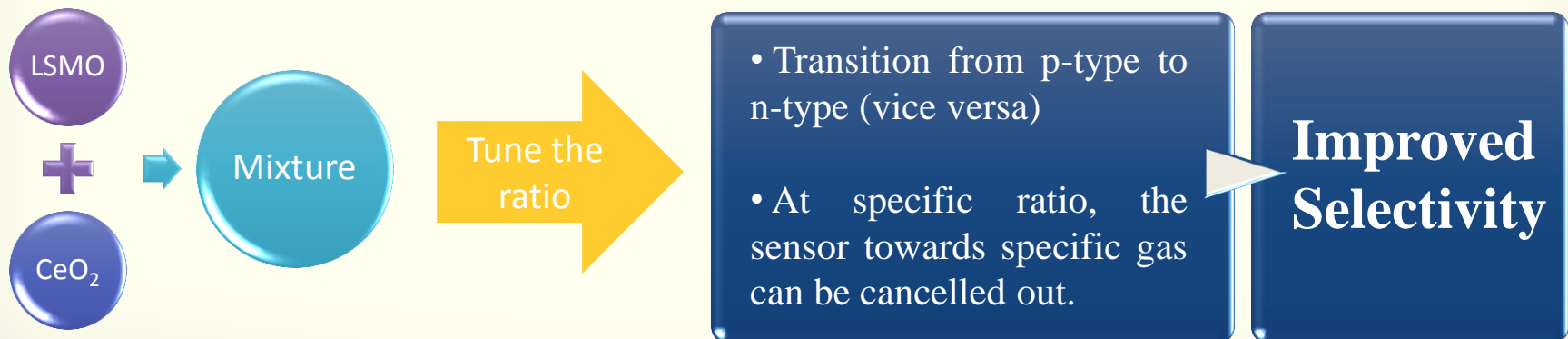
II. Engineering sensing materials for enhanced selectivity

p-LSMO/n-CeO₂ NFs heterojunction Based Sensor

❖ Properties of LSMO (p-type) and CeO₂ (n-type)

| Material | Type | Charge carrier | Resistance | In reducing gas | In oxidizing gas |
|------------------|--------|----------------|------------|-----------------|------------------|
| LSMO | p-type | Holes | Low | R ↑ | R ↓ |
| CeO ₂ | n-type | Electrons | High | R ↓ | R ↑ |

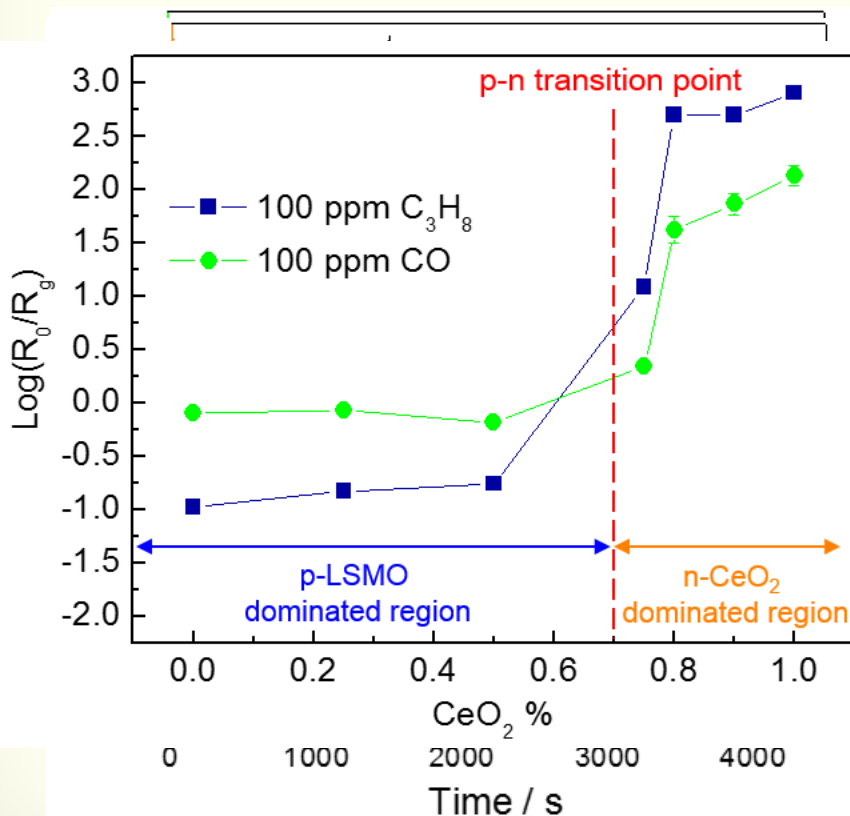
❖ Rationale



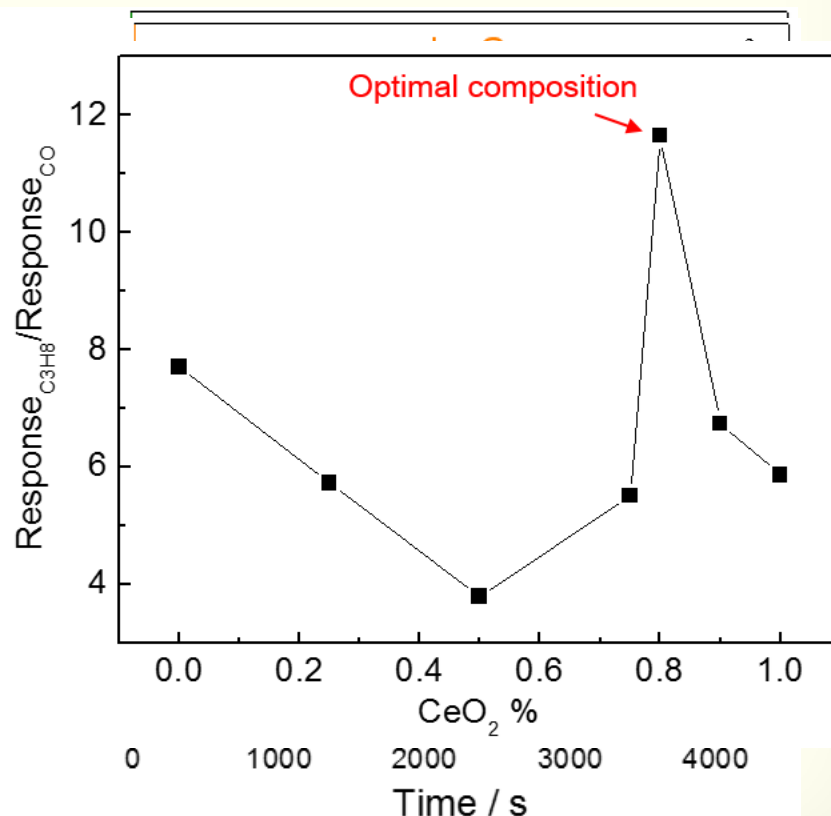
II. Engineering sensing materials for enhanced selectivity

p-LSMO/n-CeO₂ NFs heterojunction Based Sensor

CO detection

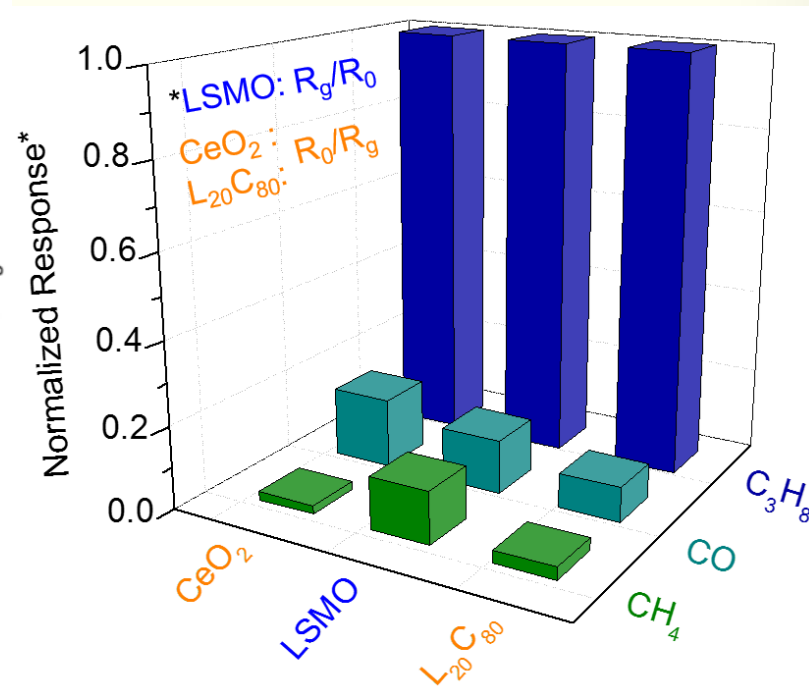
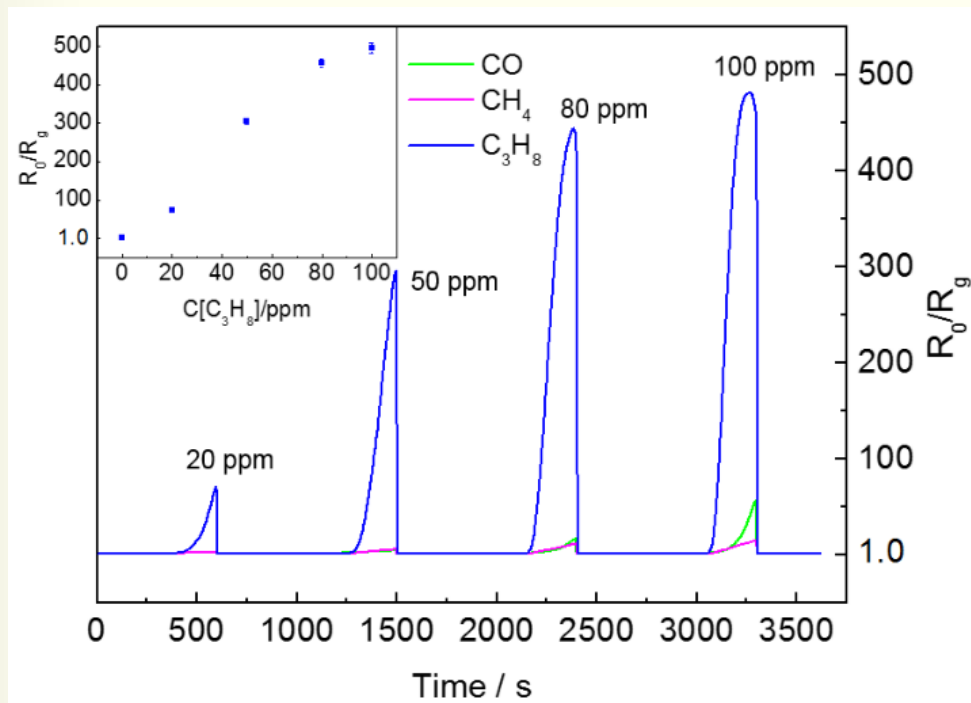


Propane detection



II. Engineering sensing materials for enhanced selectivity

p-LSMO/n-CeO₂ NFs heterojunction Based Sensor



Real-Time Detection on L₂₀C₈₀ – Based Sensor

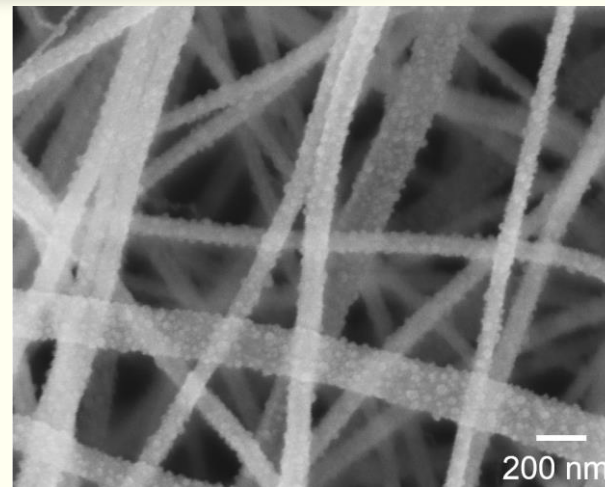
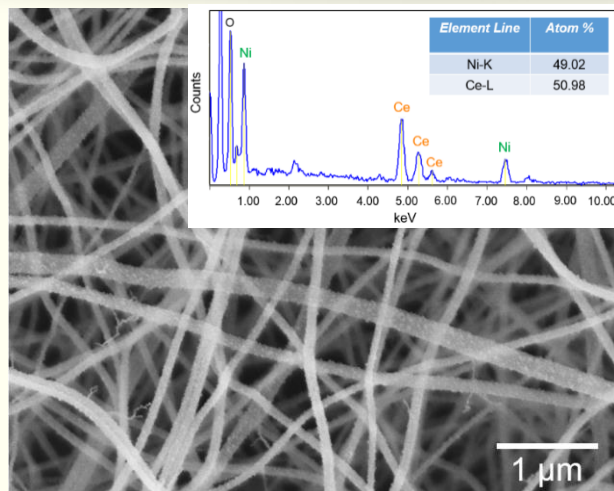


II. Engineering sensing materials for enhanced selectivity

Ce-Ni-O NFs Based Sensor

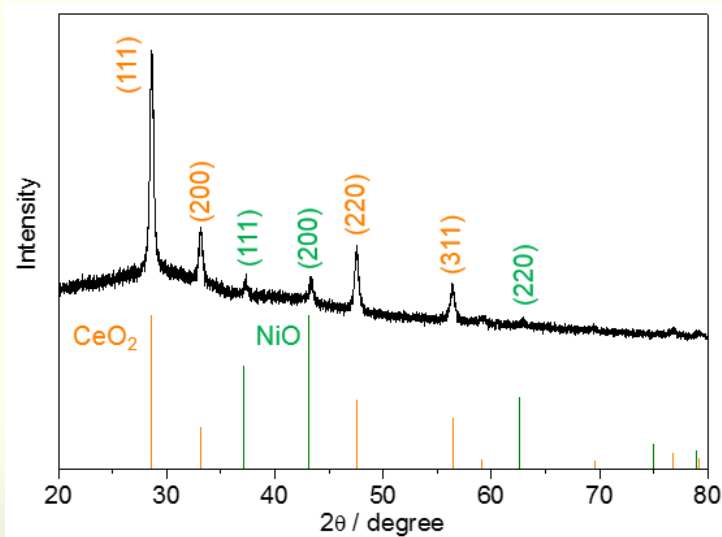
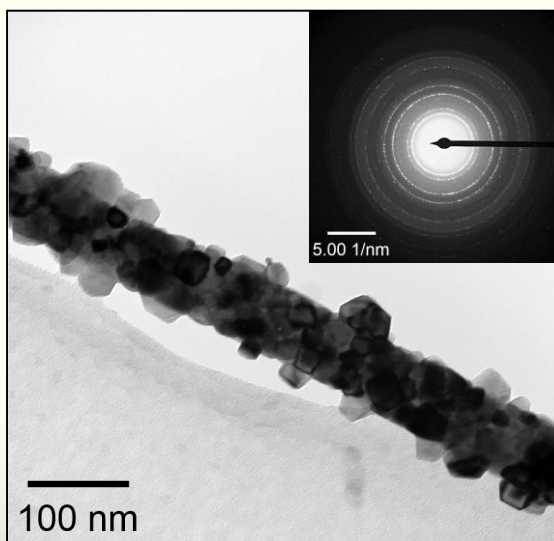
Nanofiber
 $d \sim 89 \pm 10 \text{ nm}$

Nanobelt
 $d \sim 176 \text{ nm}$



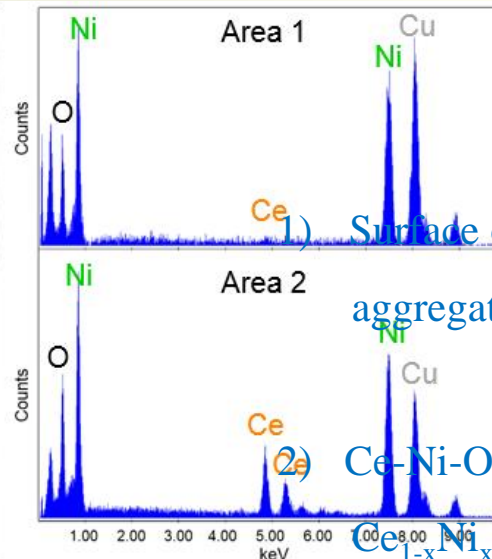
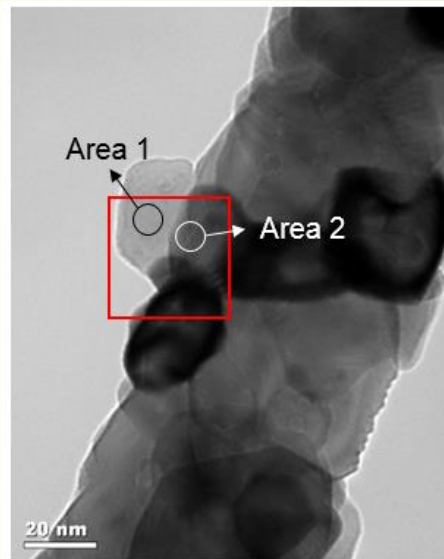
Nanofiber
 $d \sim 80 \text{ nm}$

Nanoparticles
 $d \sim 25 \pm 7 \text{ nm}$



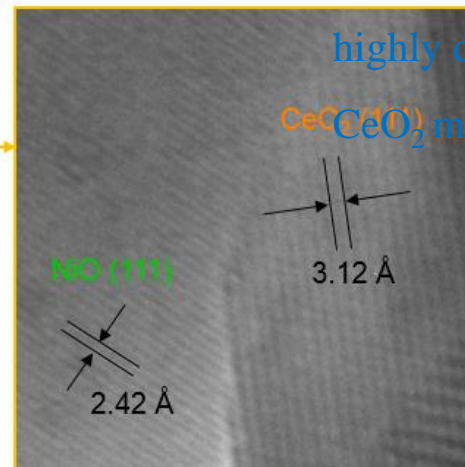
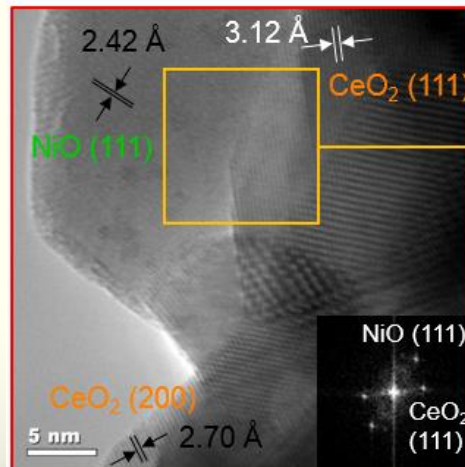
II. Engineering sensing materials for enhanced selectivity

Ce-Ni-O NFs Based Sensor



Surface of nanofibers:
aggregated NiO nanoparticles

Ce-Ni-O backbone:
Ce_{1-x}Ni_xO₂ solid solution



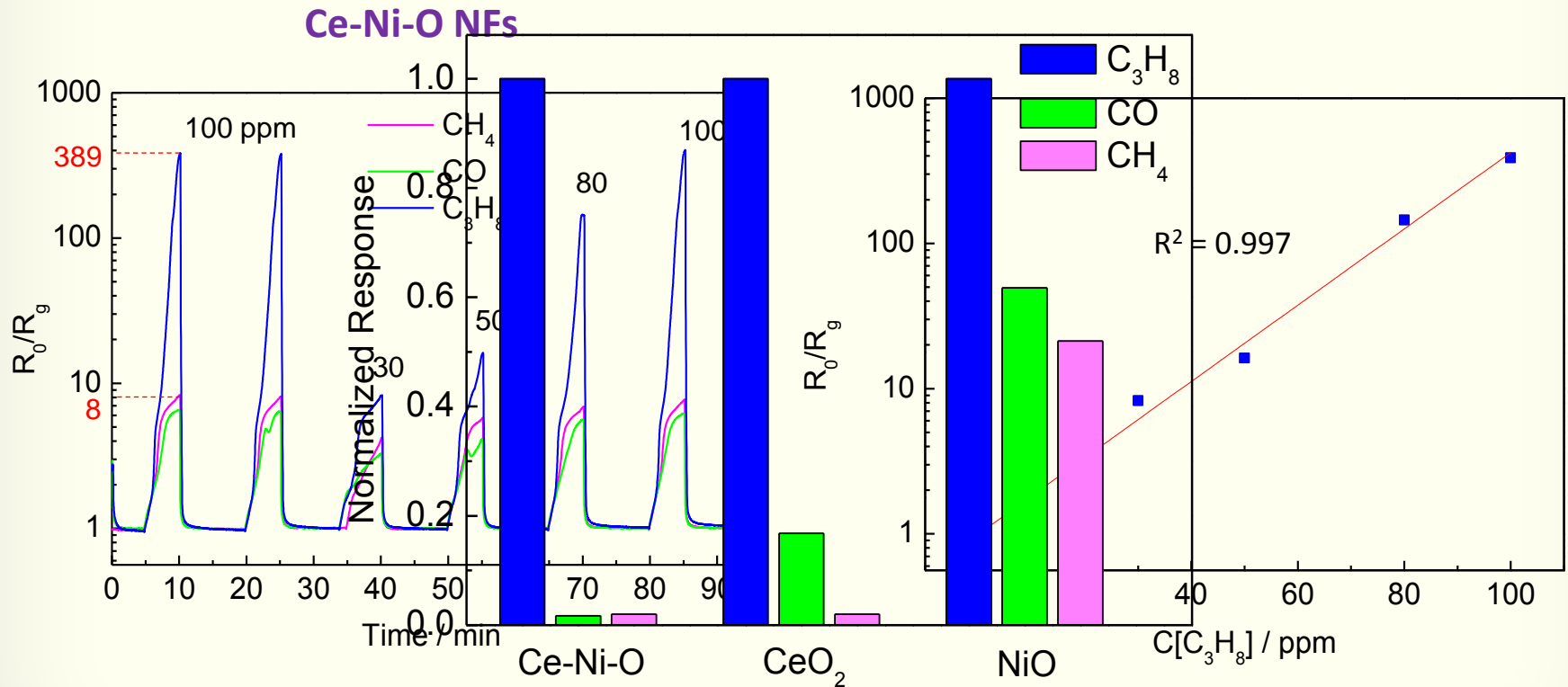
highly dispersed NiO

CeO₂ matrix



II. Engineering sensing materials for enhanced selectivity

Ce-Ni-O NFs Based Sensor



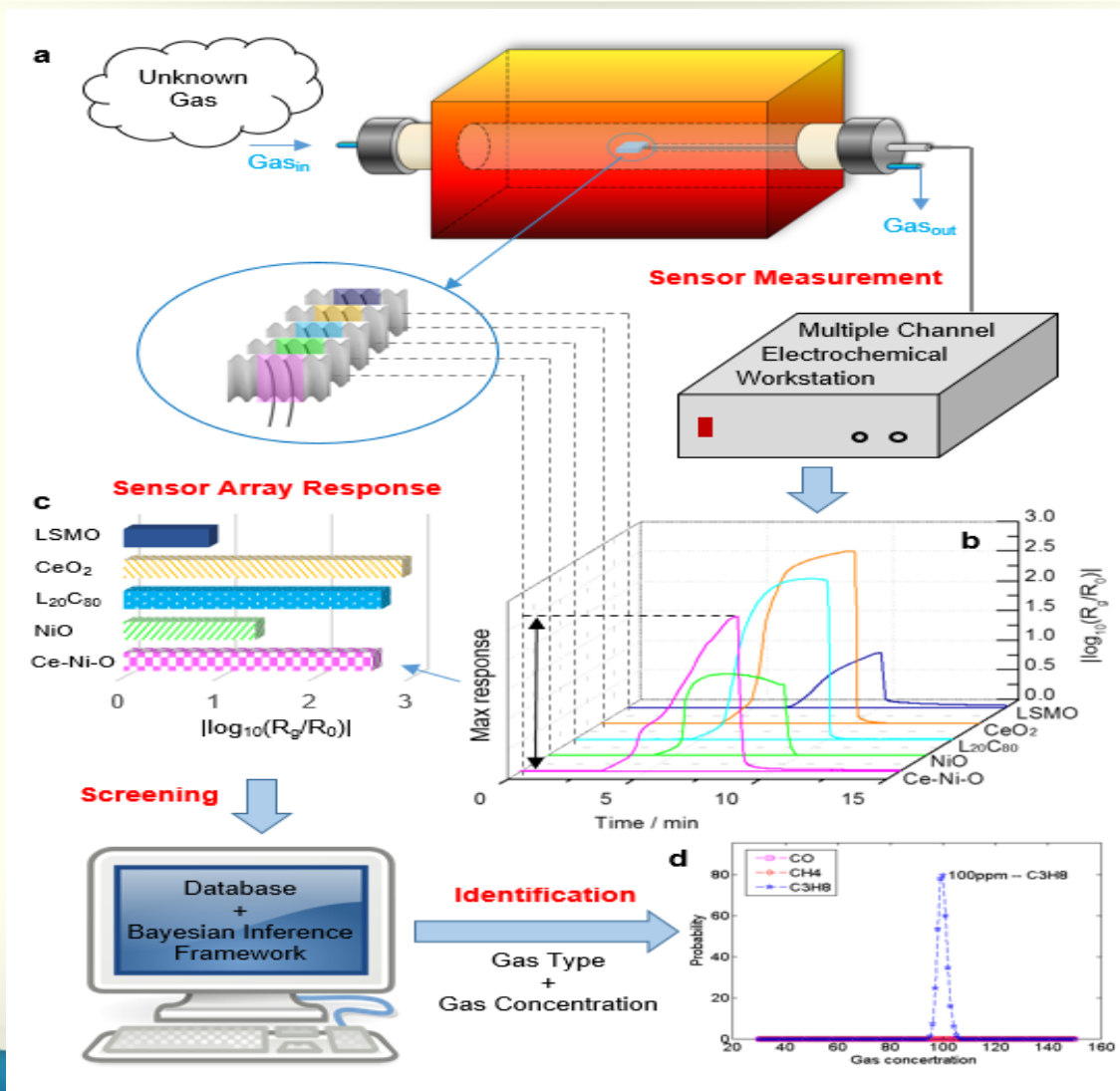
III. Developing machine learning algorithms to identify the gas concentration and species

Why to develop advanced machine learning algorithms for improved identification of target species and concentration?

- A sensor array could provide a specific and unique response patterns (fingerprints) for different individual chemical species or mixtures of species.
- With subsequent data analysis, the gas sensor array could be used to qualitatively identify gas species using pattern recognition approaches and quantitatively determine gas composition based on learning and regression methods.
- We can think of a learner as an entity that tries to guess a concept or function. Let this function be $f(x_1, x_2, \dots, x_n)$ where x_1, x_2, \dots, x_n are the underlying variables. A learner is supplied with examples. An example is nothing but a specific assignment of values to the variables and the corresponding value of the function.
- After having seen a sufficient number of examples, the learner comes up with an estimate of the function. In general the estimate may not be the same as f but possibly a close approximation to f .

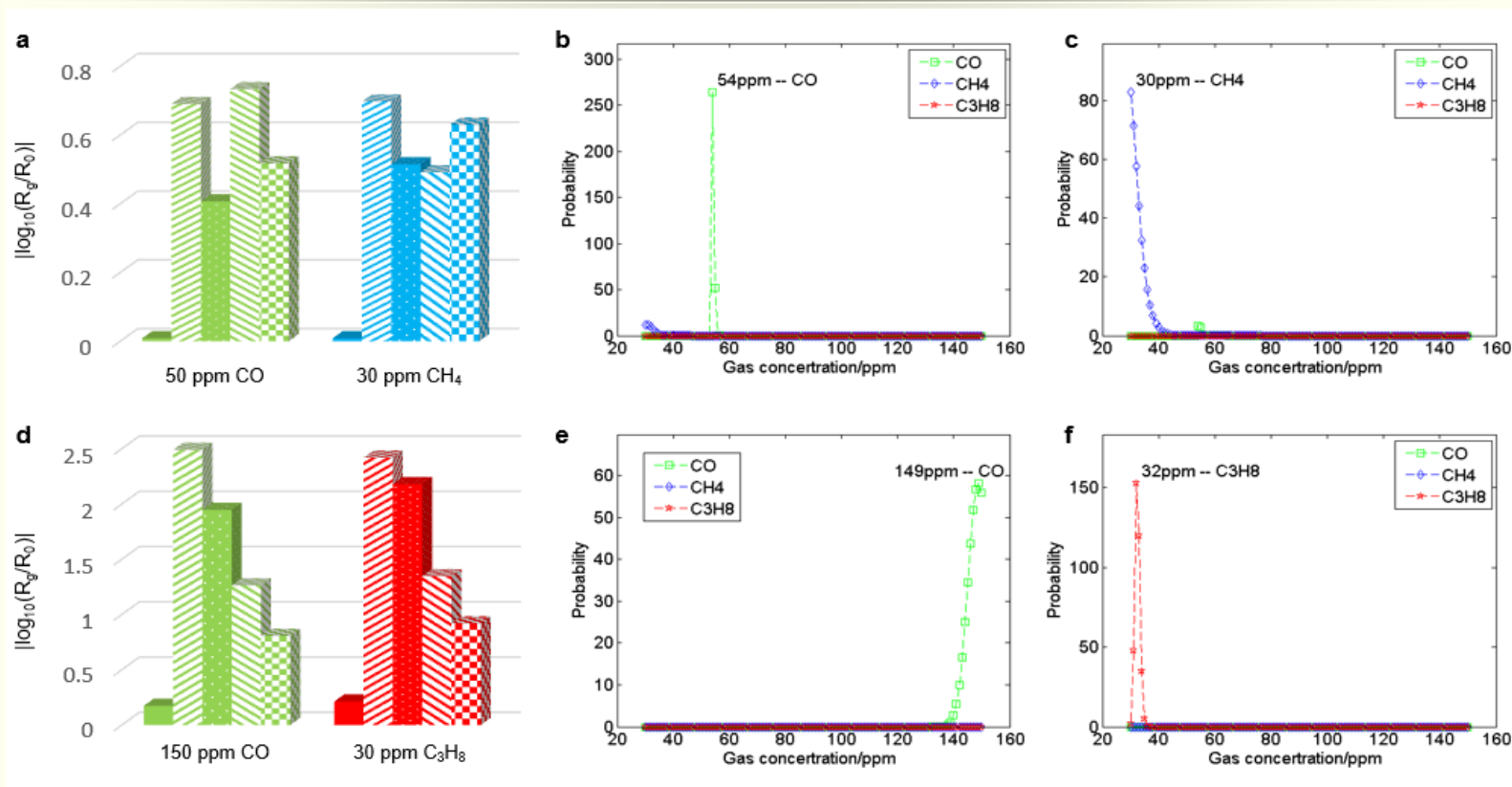


III. Developing machine learning algorithms to identify the gas concentration and species



* Yixin Liu, et al., *J. Sensors*, 2015.

III. Developing machine learning algorithms to identify the gas concentration and species



The comparison of sensor array response input of a) 50 ppm CO and 30 ppm CH₄, d) 150 ppm CO and 30 ppm C₃H₈. The output of gas identification program for b) 50 ppm CO, c) 30 ppm CH₄, e) 150 ppm CO and f) 30 ppm C₃H₈, respectively.



III. Developing machine learning algorithms to identify the gas concentration and species

Gas concentration predicted by gas identification program

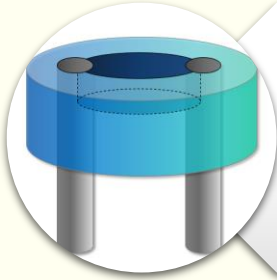
| Gas Type | CO | | | | CH ₄ | | | | C ₃ H ₈ | | | |
|---------------------|----|----|-----|-----|-----------------|----|----|-----|-------------------------------|----|----|-----|
| | 50 | 80 | 100 | 150 | 30 | 50 | 80 | 100 | 30 | 50 | 80 | 100 |
| Real Conc./ppm | 50 | 80 | 100 | 150 | 30 | 50 | 80 | 100 | 30 | 50 | 80 | 100 |
| Predicted Conc./ppm | 54 | 72 | 104 | 149 | 30 | 51 | 77 | 101 | 32 | 45 | 81 | 100 |

*Conc.: gas concentration.

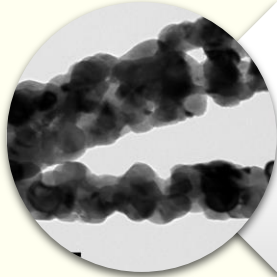


Conclusions and Outlook

Conclusions



High temperature gas sensors working in harsh exhaust environment are of paramount importance to improve combustion efficiency and control emissions.



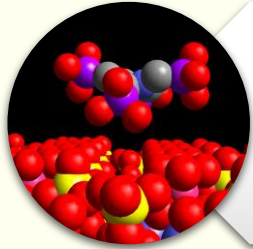
Compared to **stability** and **sensitivity**, **selectivity** is a major issue for high temperature gas sensing.



Three strategies (sensing technology, sensing materials and machine learning algorithm) have been employed to improve the high temperature gas sensing selectivity.

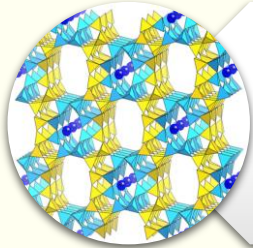
Conclusions and Outlook

Outlook



Fundamental studies for better understanding of sensing mechanisms and the relation between sensing element and sensing properties.

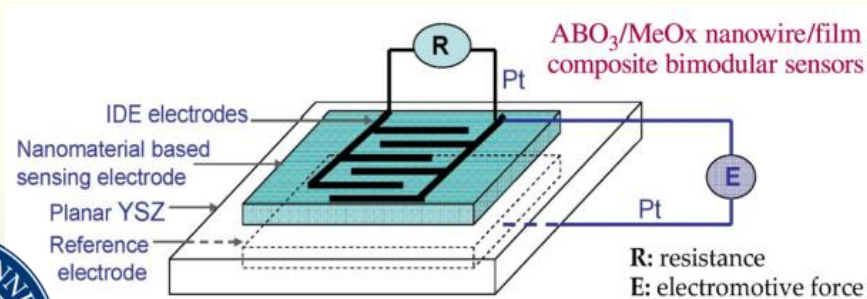
- Modeling and Simulation



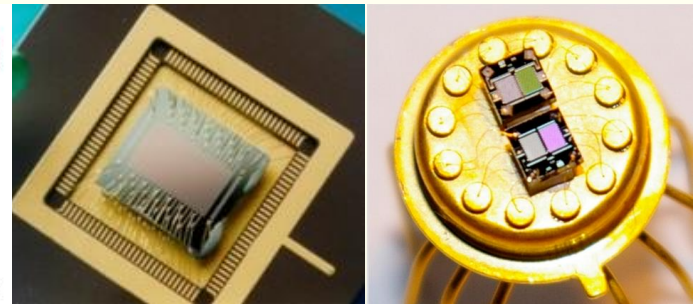
New or improved materials synthesis, novel sensor designs, combination of sensing technologies and advanced device fabrication techniques.

- Sensor array, bi-modular sensor, catalytic/physical filter

Bi-modular Sensor



Sensor Arrays



Acknowledgement

- ❖ Collaborators: Professors Pu-Xian Gao and Sanguthevar Rajasekaran at UConn
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Thank You !

Questions ?