

TOMOGRAPHIC AND HYPERSPECTRAL ANALYSIS OF POROUS THREE-DIMENSIONAL SOLID OXIDE FUEL CELL CATHODES AT MULTIPLE LENGTH SCALES

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Introduction to Solid oxide fuel cells

- Solid oxide fu
 - Low cost, flex
- Problems?
 - High operatir
 - Cathode r
 - H₂O, CO₂,
 - Adverse e
- What is primary cause
 - <u>Microstructure!</u>
 - Previous work:
 - Quantification in the FIB/SEM:
 - J. Wilson, S. Barnett, Electrochem. Commun., 11(5), 1052 (2009).
 - D. Gostovic, E. Wachsman, et al., J. Am. Ceram. Soc., 94(2), 620 (2011).

Our task:

Use the FIB/SEM (and TEM) to characterize microstructural changes as cathode degradation occurs, and relate these changes to those in cell performance.

Our goal:

Better understanding of the fundamental mechanisms behind cathode degradation.



Outline

Data acquisition

Sample prep and imaging conditions

- FIB/SEM Data processing and quantification
 Filters, artefact correction, and segmentation
 Microstructural quantification
- •STEM-EELS Hyperspectral imaging •Machine learning signal decomposition

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Experimental - Button cell testing





Cross-section view

- Symmetric cathode cells
 - 8-YSZ electrolyte
 - 50 wt. % LSM/YSZ cathode paste
- Sintered at 1000°C for 1hr
- Aged for 250hr at 800°C
 - Polarization was constant 60mA/cm²
- Four conditions compared:
 - Aged dry air
 - Aged dry air cathodic polarization
 - Aged 3% H₂O anodic polarization
 - Aged 3% H₂O cathodic polarization



FIB/SEM DATA ACQUISITION



FIB/SEM Data acquisition

• Our results (and conclusions) can only be so good as our inputs

- We need good inputs! (GIGO)
- Important considerations:
 - Initial sample preparation (pre-FIB)
 - Sample preparation within the FIB/SEM
 - Slicing resolution (for fidelity of reconstruction)
 - Electron beam parameters image noise and resolution vs. data acquisition time
 - What is it we need to accentuate?



Pre-FIB sample prep

- 1. Vacuum impregnation of porous structure
- 2. Grinding/polishing to 1200 grit
- 3. Carbon coating and sample mounting

Instrumentation

- FEI Helios 650
 - Part of the Center for Nanoscale Science and Technology (CNST) user facility at NIST
 - Multichem, iFast Developer Kit, etc.
- Auto Slice and View version 1.2
- Avizo Fire + personal Python code
- Tescan Gaia (+ Xeia) at UMD
 - Process development underway





FEI Helios 650 at NIST (CNST)



FIB/SEM DATA PROCESSING



Experimental – post processing of data

- Post-processing done with mix of software:
 - ImageJ/Python
 - Intensity gradient correction
 - Fiducial tracking/slice thickness measurement
 - Avizo Fire:
 - Non-local means filtering of data¹ (also Perona–Malik diffusion filter)
 - Watershed segmentation algorithm²

¹ Based on A. Buades *et al.* in *2005 IEEE Comput. Soc. Conf. Comput. Vis. Pattern Recognit.*, Vol. 2, p. 60. IEEE.



NL-Means Filtering





² L. Vincent and P. Soille, IEEE Trans. Pattern Anal. Mach. Intell., 13(6), 583 (1991).



Visualizations of LSM/YSZ Fuel Cell reconstruction





Visualizations of LSM/YSZ Fuel Cell reconstruction





Results – Phase percolation

- Comparison of longest section of each phase to overall network
 - Used 5 longest components
- Result:
 - YSZ and pore completely interconnected
 - LSM is limiting transport





Triple phase boundary (L_{TPB}) determination

- Intersection of three phases is necessary for the oxygen reduction reaction to occur:
 - ORR: $\frac{1}{2}O_2 + 2e^- \leftrightarrow O^{2-}$
 - This quantity can be directly related to cell performance
- Within analysis volume, a phase and boundary site can be described as **active**, **inactive**, or **unknown**
- Labels depend on connection to edges
 - Unknown have at least 1 border with edges (dead-end)
 - Active have two borders across a dimension (transverse)
 - Inactive networks have no intersection with an edge (isolated)
- Collaboration with Scientific Applications and Visualization Group at NIST
 - Implemented edge-counting more accurate than morphological expansion (current trend in literature)





Results – Triple phase boundaries



- Total ρ_{TPB} relatively constant (except H₂O-anodic, which has low sampling volume)
- H₂O-cathodic has significant decrease in active TPB density, suggesting drop in active sites for ORR



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TEM-EELS HYPERSPECTRAL IMAGING

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Electron Energy Loss Spectroscopy (EELS)

• E-beam has many interactions with specimen:



Adapted from Williams and Carter, Transmission Electron Microscopy, (2009).

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Electron Energy Loss Spectroscopy (EELS)

• E-beam has many interactions with specimen:



Adapted from Williams and Carter, Transmission Electron Microscopy, (2009).

• EELS is measurement of the energy lost when an electron interacts with the sample:



Williams and Carter, Transmission Electron Microscopy, (2009).



Decomposition analysis

- Machine learning for hyperspectral decomposition
 - How to tease out convoluted and complex signals
 - Use redundancy of information in spatial dimensions to learn more about differences in the signal dimension(s)
 - Used in EEG, audio processing, fMRI, etc.

- Non-negative matrix factorization and Blind source separation
 - Finding simpler descriptive basis vectors of overall data; one component per "source"



Adapted from: https://upload.wikimedia.org/wikipedia/commons/f/f9/NMF.png



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Hyperspectral decomposition of oxygen signal





x axis (nm)





Hyperspectral decomposition of oxygen signal

- Discrete surface component
 - Independent from LSM or YSZ
- Lack of peak in O-K EELS signal at surface
 - Suggests decrease in oxygen 2nd nearest neighbors
 - More oxygen vacancies
- Increase in edge onset (compared to LSM)
 - Increased bandgap, lower electrical conductivity



D. Muller, et al., Nature., 399, 758 (1999).



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 f_{Mn}

Hyperspectral decomposition – Mn segregation

Mn L_{2.3} EELS Spectrum Image 70 nm

x axis (nm)



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Mn-segregation valence analysis





Mn-segregation valence analysis



S.J. Shih, et al., J. Electrochem. Soc., 158(10), B1276, (2011).



Mn-segregation valence analysis



- EELS reveals mobile Mn ions with valence of \approx 2.5 at the interface of YSZ particles
 - Suggests not a full perovskite coordination



Summary

Conclusions

- Developed methods to quantify 3D microstructure of solid oxide fuel cell cathodes
- Subtle changes in LSM/YSZ microstructure occur; which agree with subtle changes in cell performance
- $\rho_{\text{TPB,active}}$ decreases when aged under H₂O contamination and cathodic polarization
- Segregation of Mn and La to YSZ grain boundaries in H₂O-cathodic (but not Sr); change in LSM Mn valence

Ongoing Work

- Analyze and quantify composition of segregation products using TEM/EELS
- Further correlation with EIS data from same samples
- Investigation of LSCF/GDC composite cathode degradation



Acknowledgements





THANK YOU

Questions and comments?

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