

Chemical and Physical Degradation in Solid Oxide Cells: Modeling and Optimization

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Motivation

- High temperature operation of solid oxide cells (SOCs) increases their efficiency but accelerates physical and chemical degradation resulting in reduced cell life.
- Non-destructive measurement of time-varying extent of degradation is practically impossible.
- Degradation models are developed for optimal operation and control under electrolysis and power modes.
- Effects of physical and chemical degradation are generally studied independently in the open literature. This study also investigates their synergistic effects.

Optimization of Long-Term Operation

- Chemical degradation results in changes in the specific energy consumption and thermal performance of the SOC.
- Dynamic optimization of long-term operating conditions is necessary to obtain efficient operation over the lifetime of the system.

Maximize Integral Efficiency over lifetime

$$\max \sum_t \eta_t = \sum_t \frac{HHV(H_2 \text{ produced}_t)}{P_{total,t}}$$

Minimize Voltage Degradation Rate

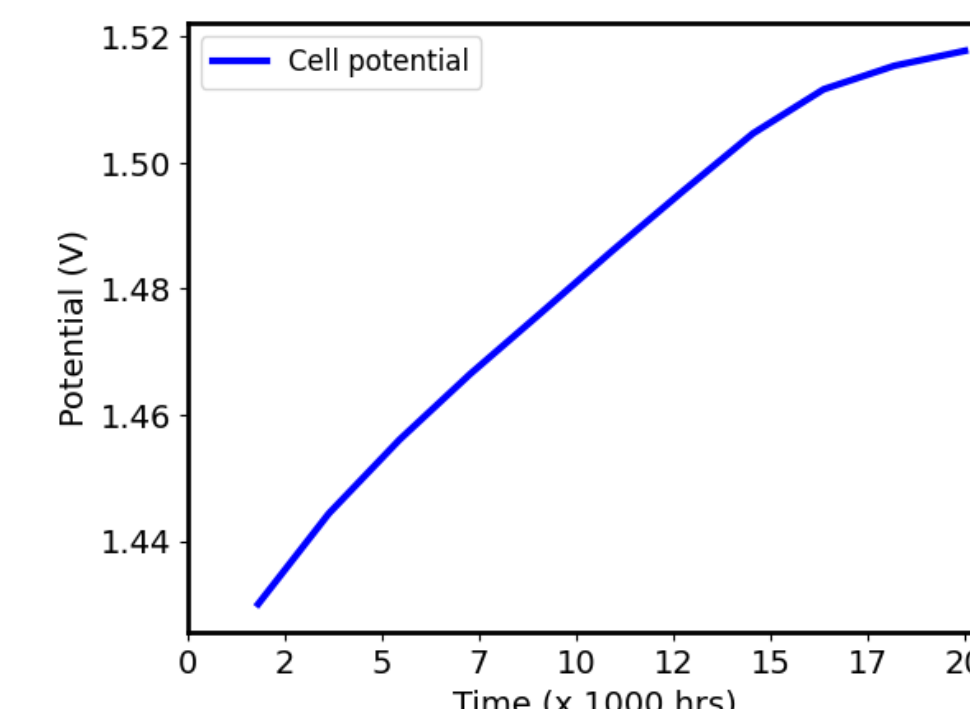
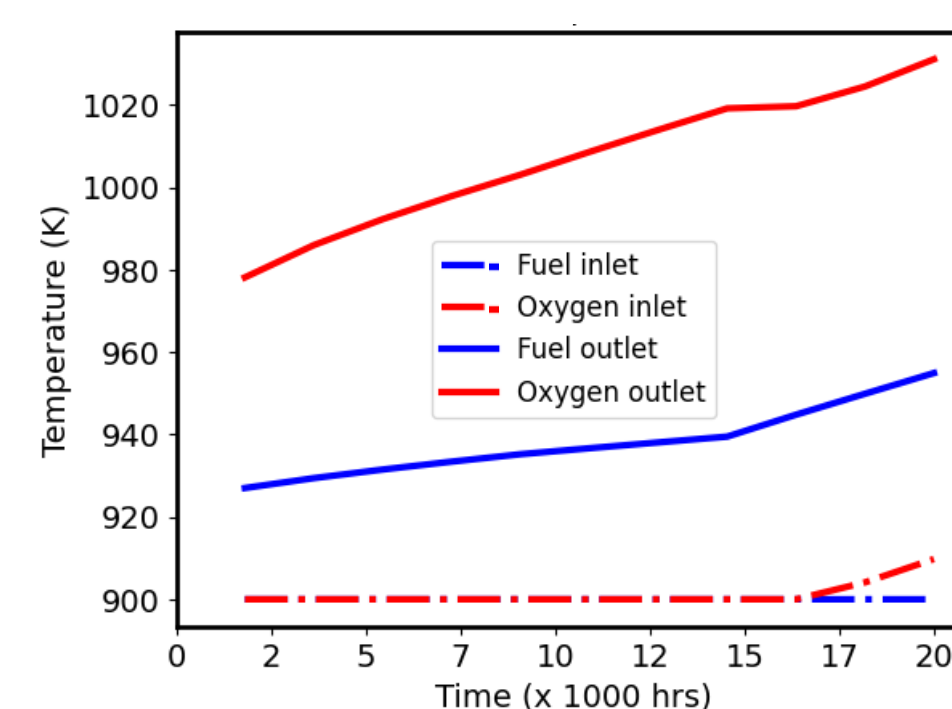
$$\min \frac{\tilde{V}_{tf} - \tilde{V}_0}{\tilde{V}_0}$$

Maximize Lifetime H₂ production

$$\max \sum_t H_2 \text{ produced}_t$$

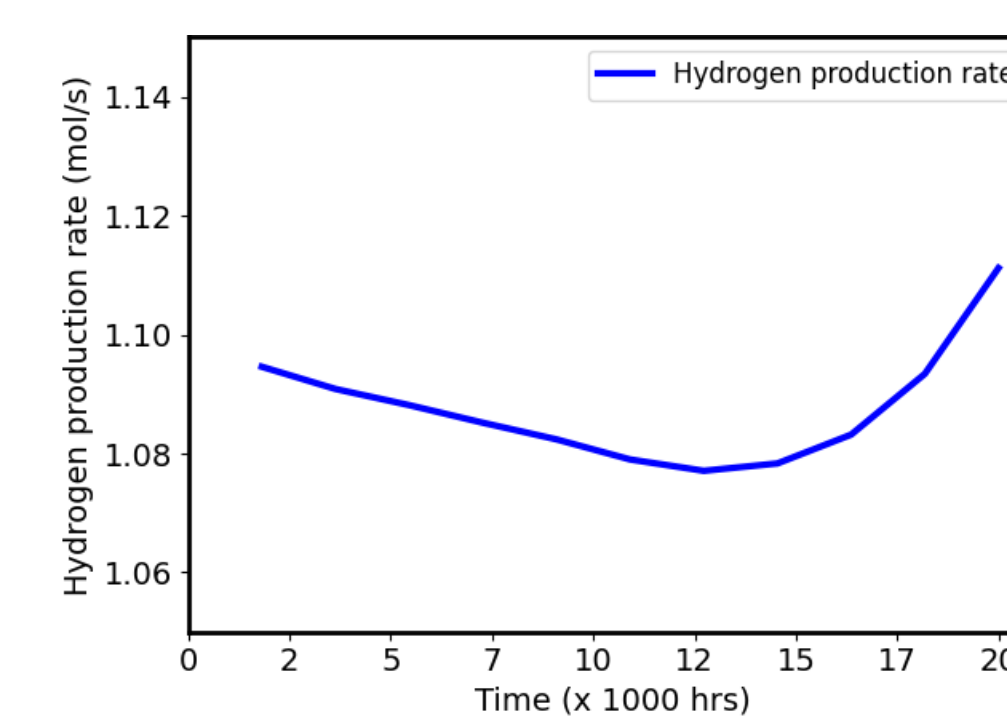
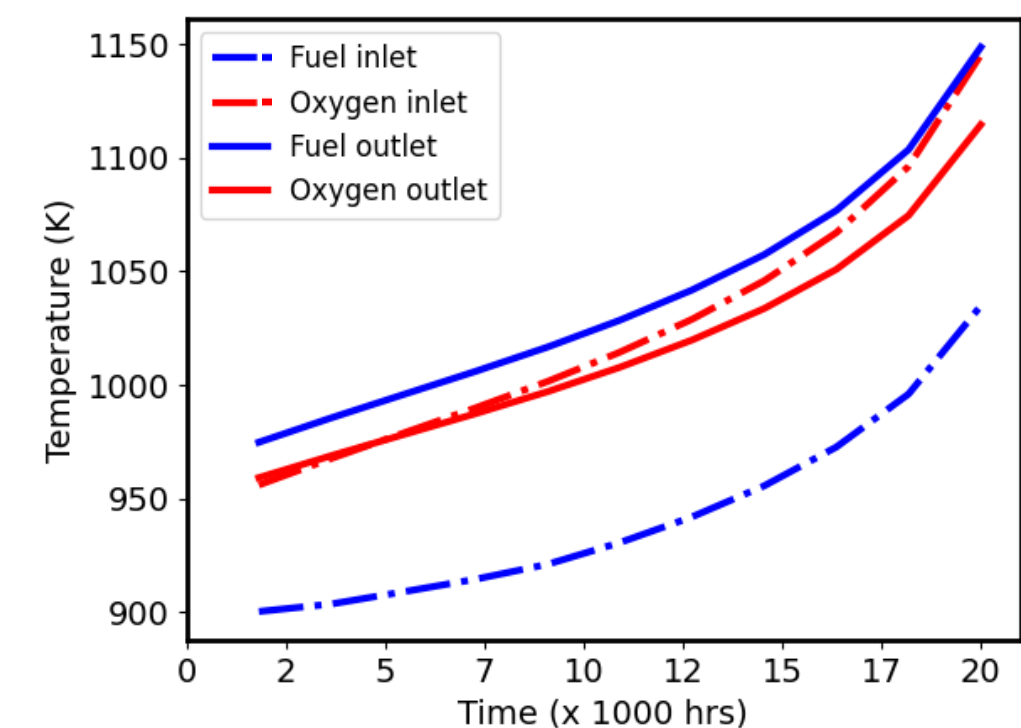
Operating Mode	Objective Function	Specific Energy Consumption (kWh/kg)	Lifetime H ₂ Production (× 10 ⁸ Kg)	Voltage Degradation Rate (%/khr)
Constant Current Density	Undegraded Cell	35.3	118	None
	Max integral efficiency	36.6	118	2.6
	Min degradation rate	40.4	118	1.8
Constant Potential	Max Integral Efficiency	35.6	88	1.7
	Max lifetime H ₂ prod.	38.9	100	2.0
	Min final degradation	41.7	40	0.7

Constant Current Density Operation – Minimize Degradation Rate

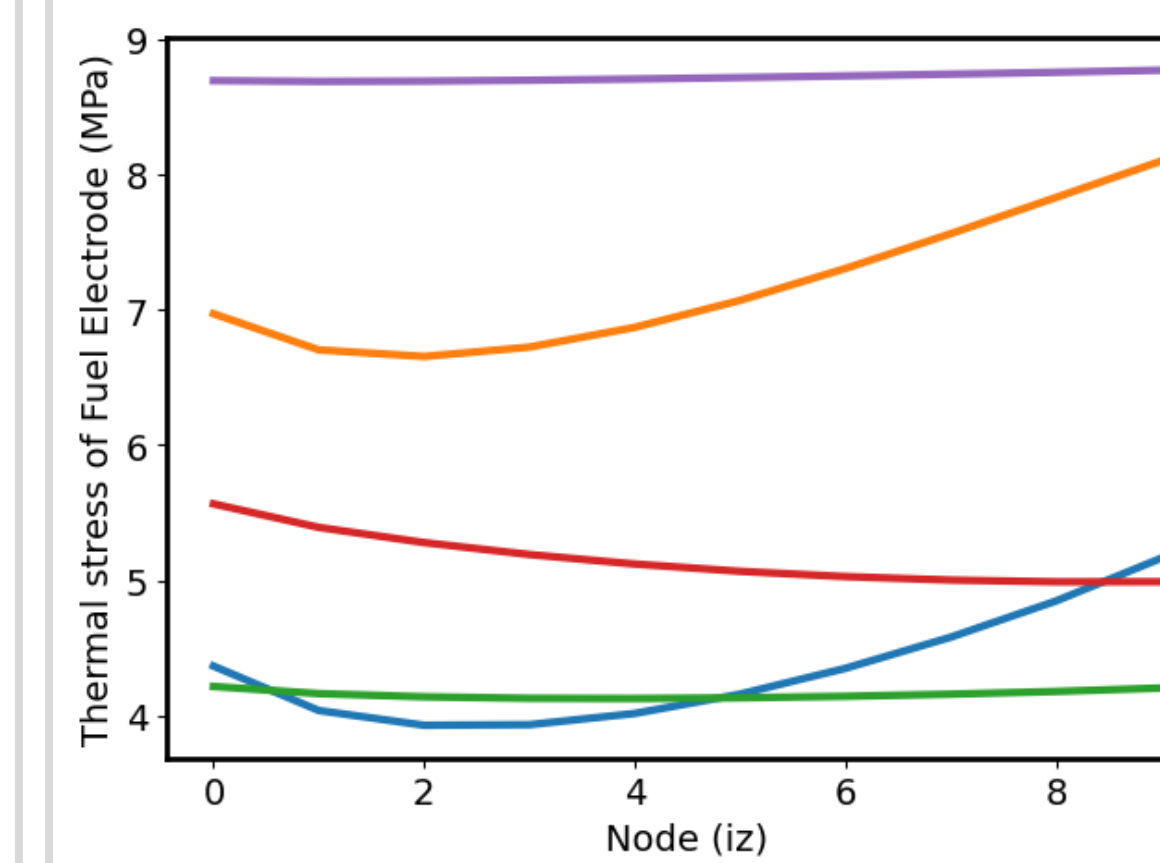


- Constant current density (constant H₂ production rate) operation results in a high degradation rate.
- It is observed that to minimize chemical degradation, it is desired to have low inlet temperatures.
- Tradeoffs between specific energy consumption and degradation rate can be captured depending on the pricing of electricity and hydrogen.

Constant Potential Operation – Maximize Integral Efficiency

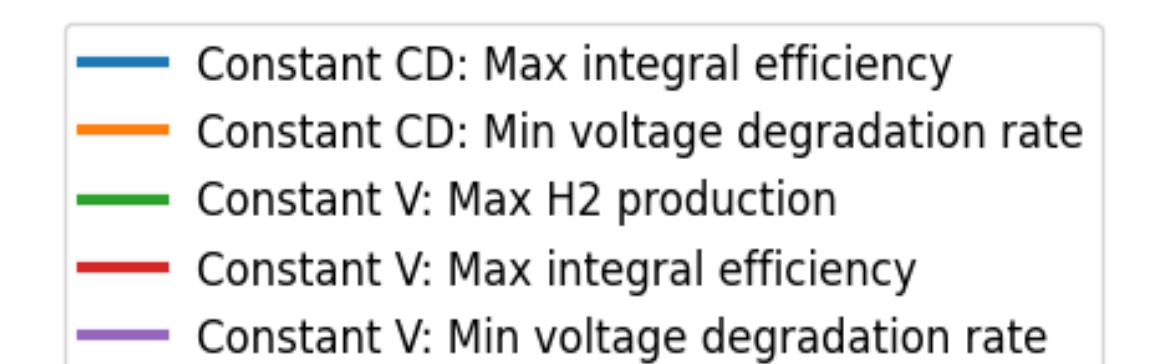


Thermal stress characteristics



Thermal Stress Profiles after 20,000 hours

- A thermal stress model calculates stresses due to change in the coefficient of thermal expansion.
- Different optimal operating scenarios have different stress distributions.
- Tensile stress in the fuel electrode can result in accelerated failure.
- Optimizing operation to minimize chemical degradation can result in increased physical degradation.

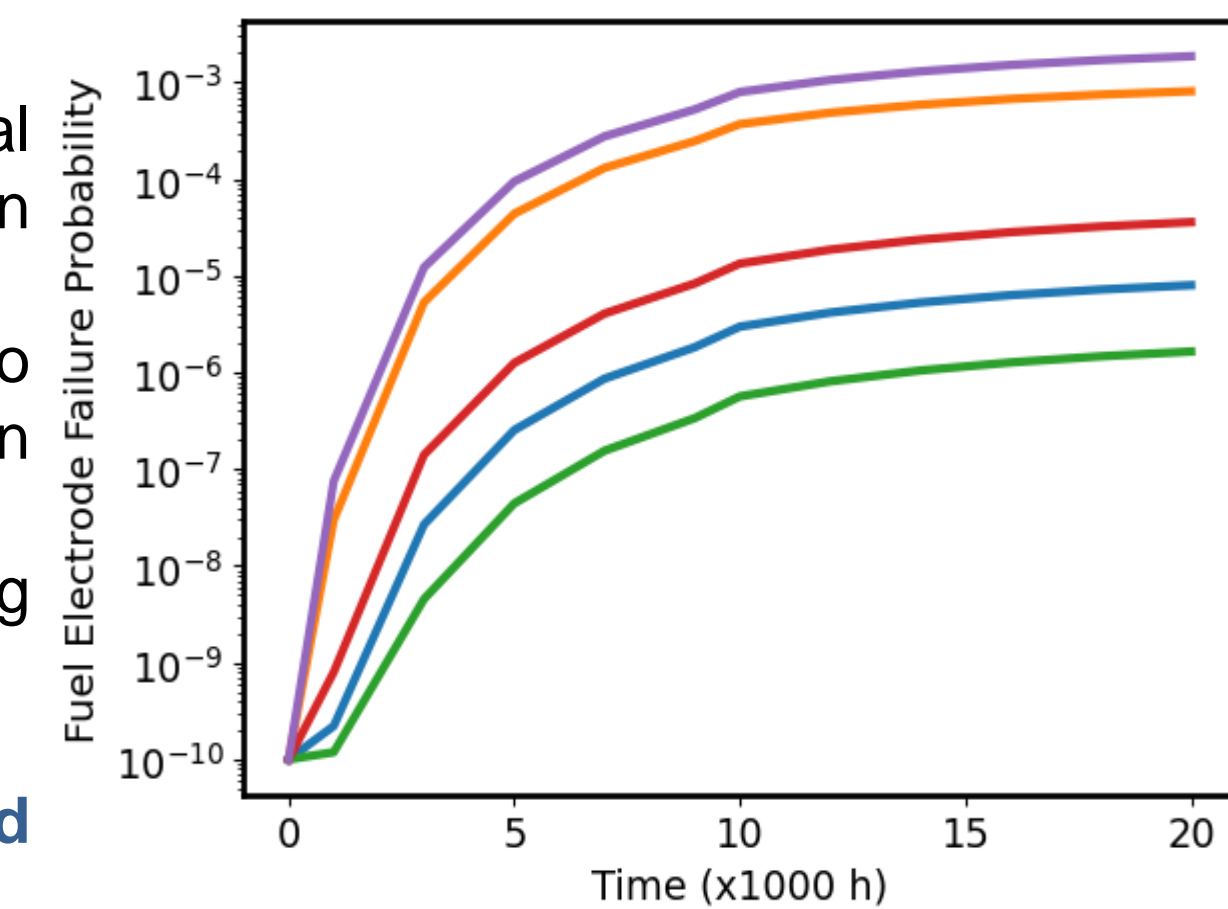


For thermal stress calculations:

- Reference stress free at $T_{ref} = 1473 \text{ K}$

Creep Induced Failure Probability

- Extended operation at high thermal stress results in increased creep strain in the SOC.
- The Weibull distribution is used to estimate the failure probability of a given layer.
- The effect of different optimal operating scenarios on failure probability is shown.



- A tradeoff exists between physical and chemical degradation phenomena.

Concluding Remarks

- Increased resistance within the cell due to chemical degradation result in higher operating temperatures.
- Time-varying changes in the internal cell temperatures can result in thermal stress buildup.
- Dynamic optimization over lifetime can identify operational strategies to minimize degradation or maintain low specific energy consumption.

Future Work

- Extend degradation models to other leading electrode and electrolyte materials.
- Undertake optimization to minimize LCOH by taking into account operating costs, hydrogen selling price, stack replacement schedule, and stack cost.
- Investigate optimal operating strategies for seasonal H₂ demand.
- Extend thermal stress and creep calculations to sealant layer.
- Develop algorithms for rigorous multi-scale optimization considering both fast and slow time scales over large time horizon.

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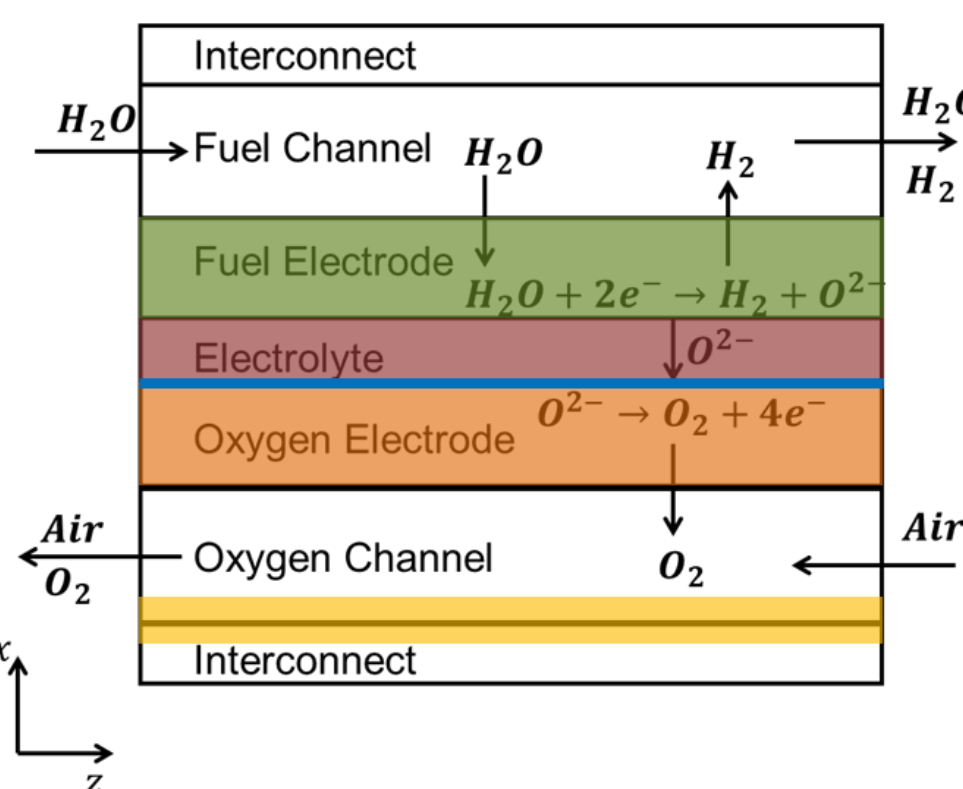
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SOC Degradation Modeling Capabilities

- Chemical degradation is caused by changes in the triple phase boundary (TPB) microstructure due to oxidation, agglomeration and redistribution that in turn affect conductivity, active area, and efficiency.
- Degradation rates are substantially higher under electrolysis due to high oxygen partial pressures in the oxygen electrode.
- Physical degradation is induced by thermal excursion and cycling and causes creep and fatigue damage in cells due to thermal stress evolution.
- Thermal stresses evolve because of time-varying differential thermal expansion between the three layers.

Chemical degradation modeling capabilities

- Dynamic first principles and empirical models include five dominant degradation mechanisms in the three active layers of the SOC.
- Models include parabolic growth laws for oxide growth, Ostwald ripening for Ni agglomeration, and Fick's law for surface diffusion.



Physical degradation modeling capabilities

- Thermal stress models are developed that can capture the effect of time-varying temperature profile within the cell.
- Creep evolution can be calculated, and failure probability analyzed.

Coupling and Solution Methodology

Oxygen electrode degradation mechanisms
Chromium oxide scale growth
Lanthanum zirconate scale growth
LSM-YSZ coarsening
Fuel electrode degradation phenomena
Ni agglomeration and volatilization
Electrolyte degradation phenomena
YSZ electrolyte delamination

- Degradation models set up as sub-models within SOC model.
- Full discretization over 20k hours of operation is intractable.
- Quasi-steady state assumption for flowsheet level dynamics.
- Model discretized according to degradation dynamics
- Solved fully discretized DAE system as a nonlinear programming problem using ipopt.

