

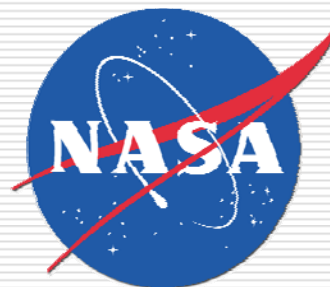
# Optimal Experiment Design for Thermal Characterization of Functionally Graded Materials

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# Outline

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- Motivation: Design experiments to measure thermal properties of functionally-graded (FG) materials.
- Simulation studies of transient heating:
  - In FG material with  $k(x)$ , conduction only;
  - In high-porosity material with large  $\Delta T$ , both conduction and radiation present.
- Summary and future work.

# Functionally-graded (FG) materials

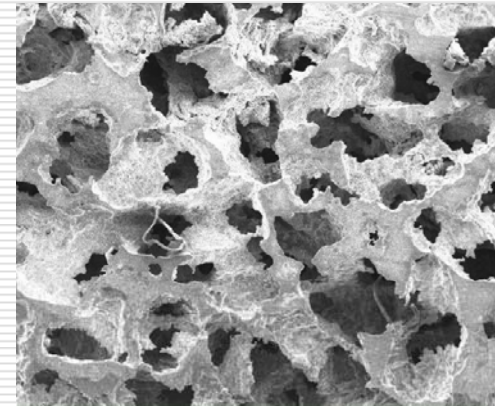
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- ❑ Material (or structure) with properties that vary through the thickness
- ❑ Property variations are of interest to improve thermal/stress performance
- ❑ FG materials include composites, built-up structures, metal foams, or any structure with variations designed into the material

# Fabrication Methods for FG Materials

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- Layer processing
  - Mechanical lamination
  - Spraying
  - Vapor deposition
- Bulk processing
  - Powder mixing and sintering
  - Fibers
- Diffusion processing
- Melt processing



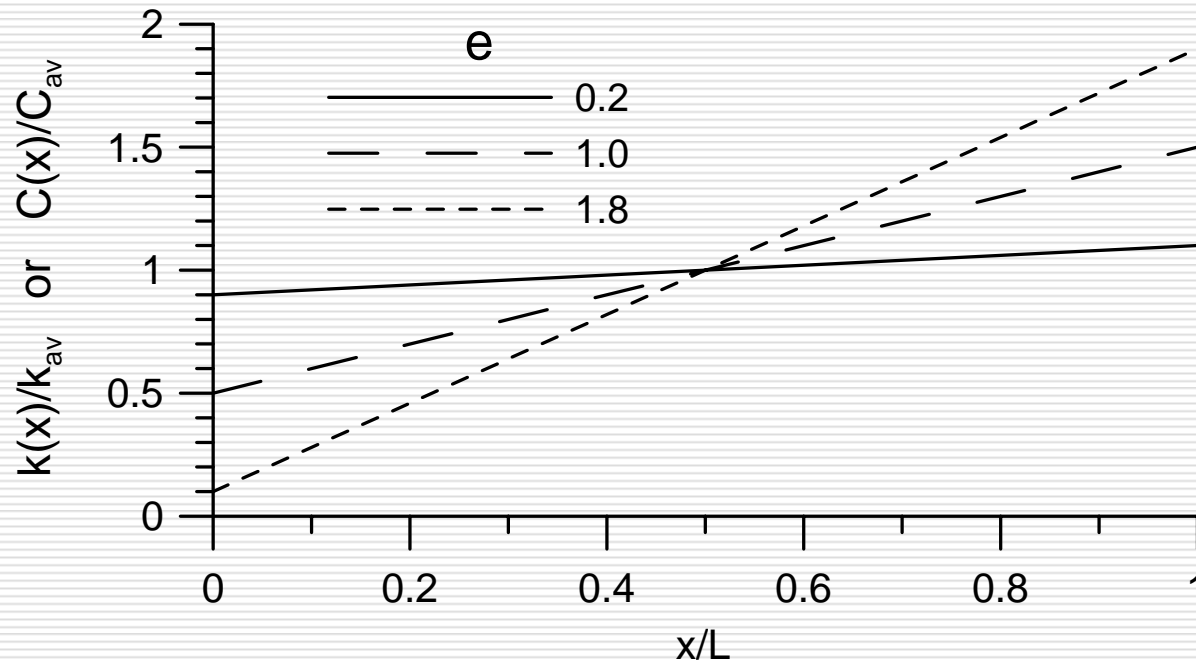
Example: porous metal foam with spatially-varying porosity.

# Motivation for current project

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- ❑ Functionally graded (FG) materials are being studied for thermal insulation (for aero-thermal protection or cryogenic tanks)
- ❑ Once fabricated, does a delivered material meet the specifications?
- ❑ Seek procedure to choose an experiment to measure spatially-varying thermal properties of FG materials (accurate; non-invasive; cost effective)

# FG Material, linear variation of $k(x)$



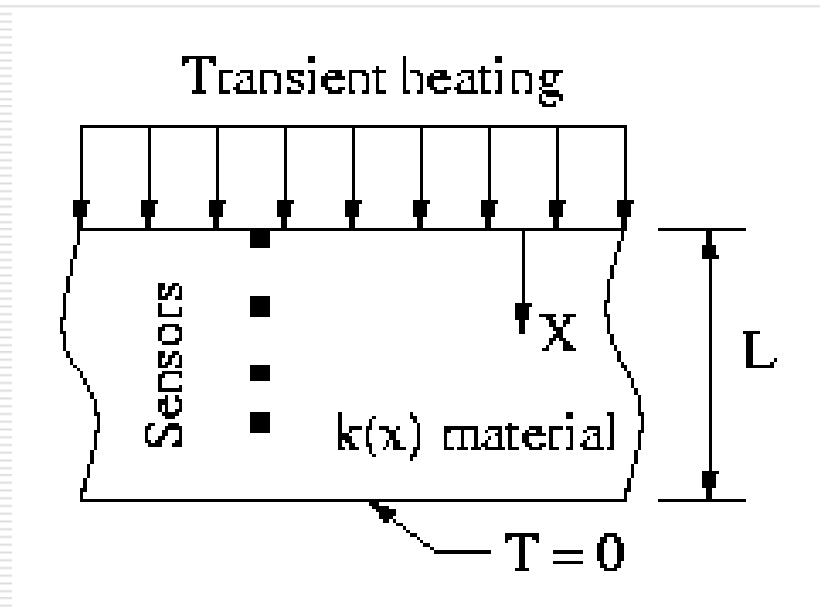
□  $k(x) = k_{av} [1 + e(x/L - 0.5)]$  (W/m/K)

□  $C(x) = C_{av} [1 + e(x/L - 0.5)]$  (J/m<sup>3</sup>/K)

# Transient experiment for thermal property measurement.

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- ❑ On-off heating on one side
- ❑ Fixed temperature on other side
- ❑ One or more temperature sensors
- ❑ Collect data during heating and continue while unheated.
- ❑ Analyze data for desired properties with parameter estimation.



# Parameter Estimation

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- Goal: given experimental data  $Y_j(\theta_i)$ , find parameters  $(k_{av}, C_{av}, e)$ .
- Need model: given  $(k_{av}, C_{av}, e)$ , compute  $T_j(\theta_i)$ . (j ... sensor; i ... time)
- Minimize the sum-of-square of error  
$$[ Y_j(\theta_i) - T_j(\theta_i) ]^2$$
- Minimization requires derivatives of the model temperatures with respect to the parameters (sensitivity coefficients).



# Computational Model

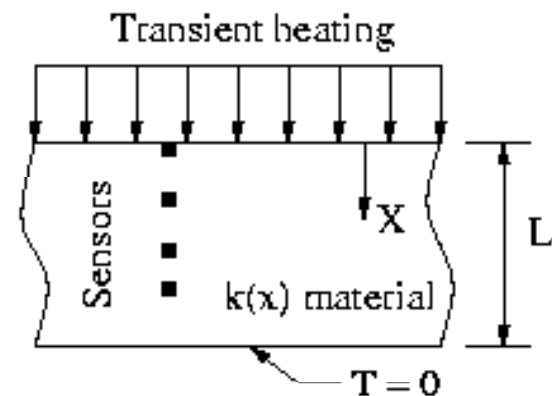
(solution method: finite difference)

$$\frac{\partial}{\partial x^+} \left( k^+ \frac{\partial T^+}{\partial x^+} \right) = C^+ \frac{\partial T^+}{\partial \theta}; \quad 0 < x^+ < 1$$

$$\text{at } x^+ = 0, \quad -k^+(0) \frac{\partial T^+}{\partial x^+} = \begin{cases} 1, & \theta < \theta_h \\ 0, & \theta > \theta_h \end{cases}$$

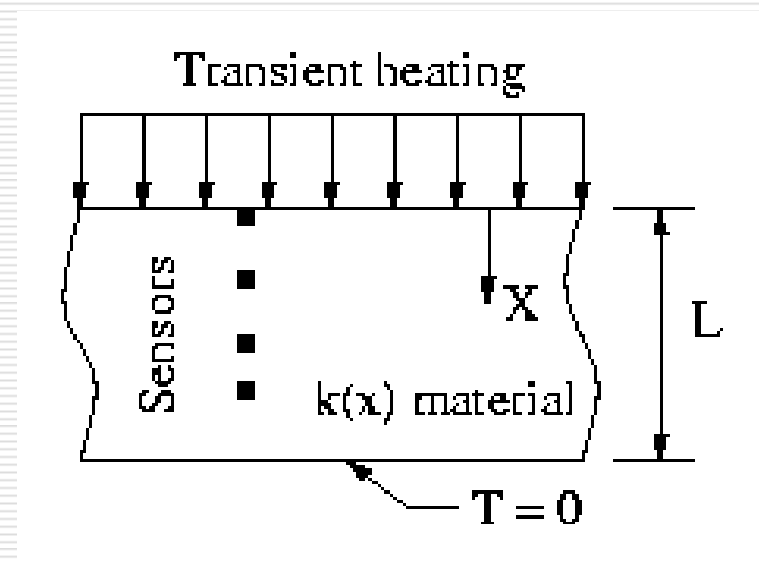
$$\text{at } x^+ = 1, \quad T^+(1, \theta) = 0$$

$$\text{at } \theta = 0, \quad T^+(x^+, 0) = 0$$



# Seek best experiment by varying:

- ❑ Number and location of sensors
- ❑ Duration of heating period
- ❑ Duration of entire data record
- ❑ Location of heating ( $x=0$  or  $x=L$ )
- ❑ Apply to several different spatial variations (slope,  $e$ )



# Compare 2 Experimental Designs

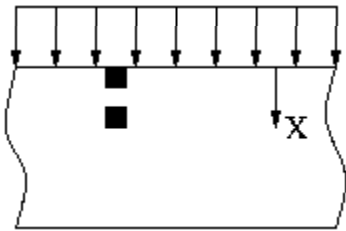
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Experiment 1:

Heat at  $x=0$

with two sensors:

- Sensor at  $x=0$ .
- Sensor at  $x=L/4$ .

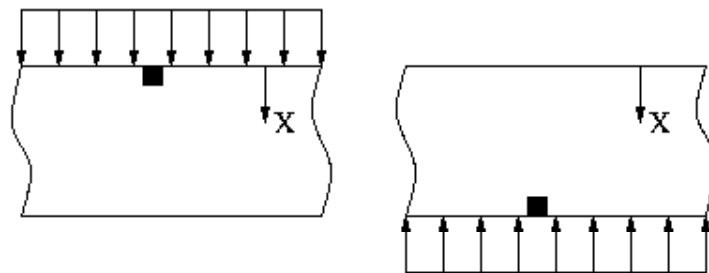


Experiment 2:

Two heating events

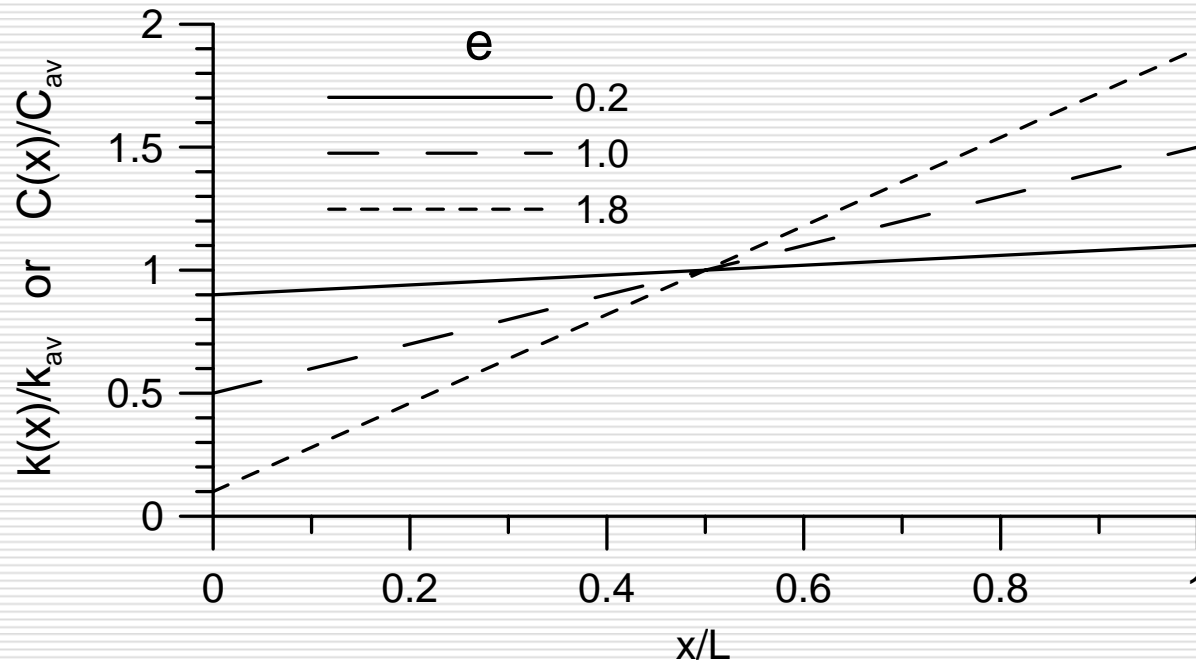
with one sensor each:

- Heat/sensor at  $x=0$ .
- Heat/sensor at  $x=L$ .



Seek optimum for each experimental design by varying heat duration and data duration, for several values of slope  $e$ .

# FG Material, linear variation of $k(x)$

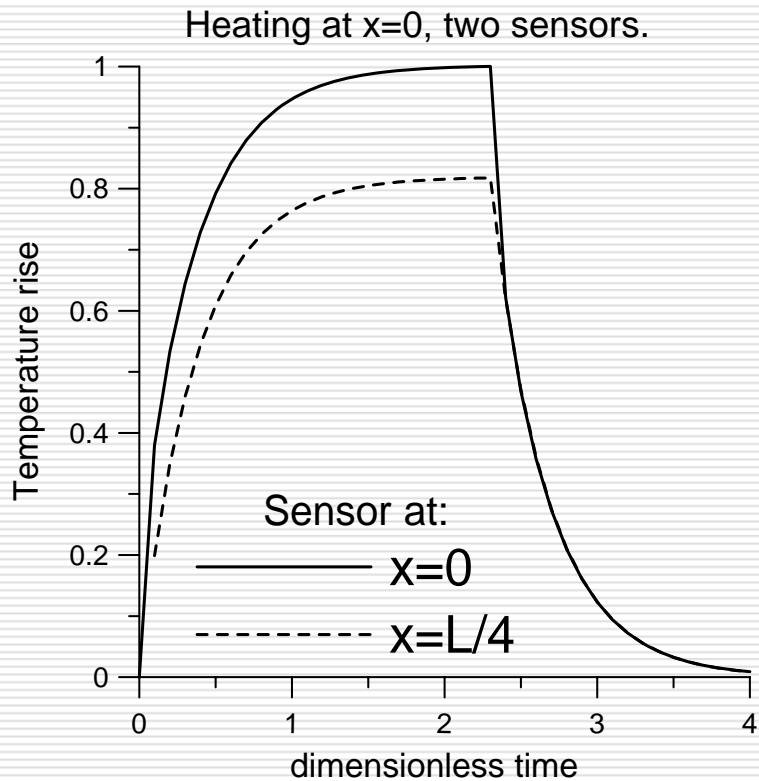


□  $k(x) = k_{av} [1 + e(x/L - 0.5)]$  (W/m/K)

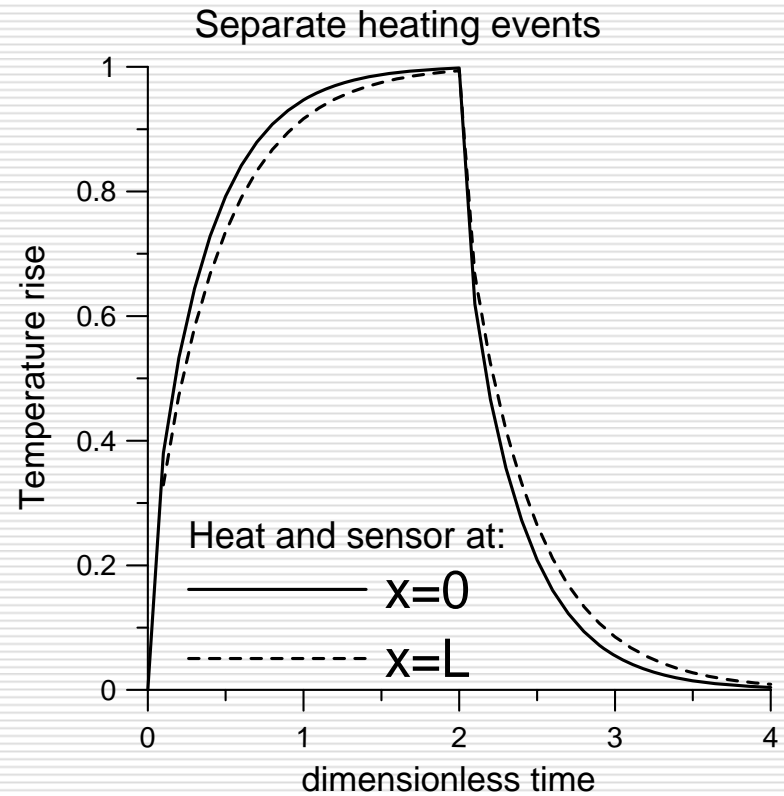
□  $C(x) = C_{av} [1 + e(x/L - 0.5)]$  (J/m<sup>3</sup>/K)

# Calculated temperatures ( $e = 0.2$ )

## Experiment 1.



## Experiment 2.



# Sensitivity coefficients

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- Minimization of error requires sensitivity coefficients (derivatives wrt parameters  $b_k$ )

$$X_{jk}(i) = b_k \frac{\partial T_j^+}{\partial b_k}$$

- which are computed from the model:

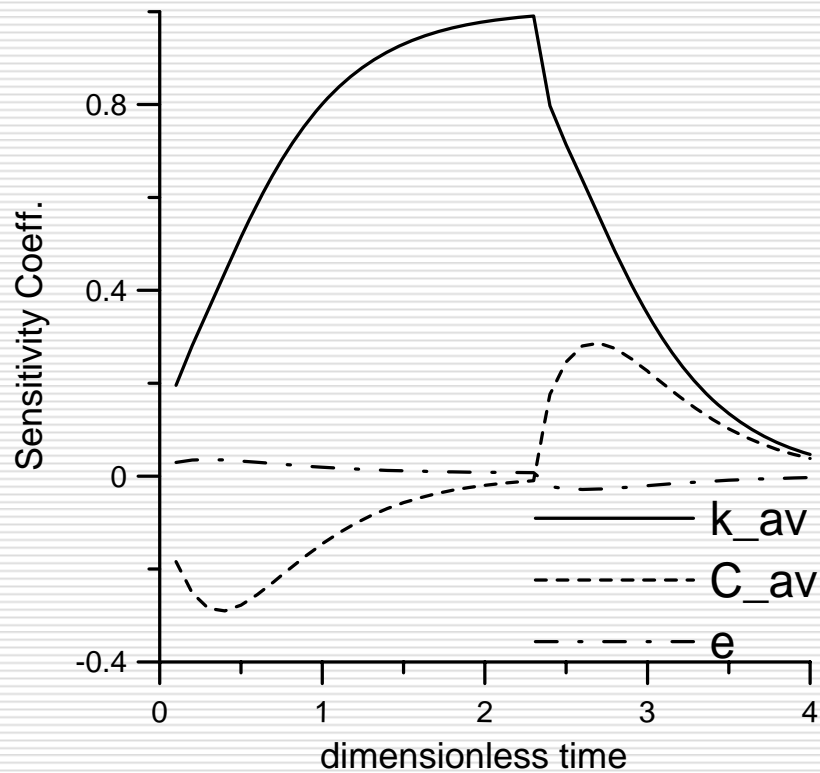
$$X_{jk}(i) \approx b_k \frac{[T_{ij}^+((1 + \epsilon)b_k) - T_{ij}^+(b_k)]}{\epsilon b_k}$$

- and assembled into matrix **X**.

# Sensitivity Coefficients ( $e = 0.2$ )

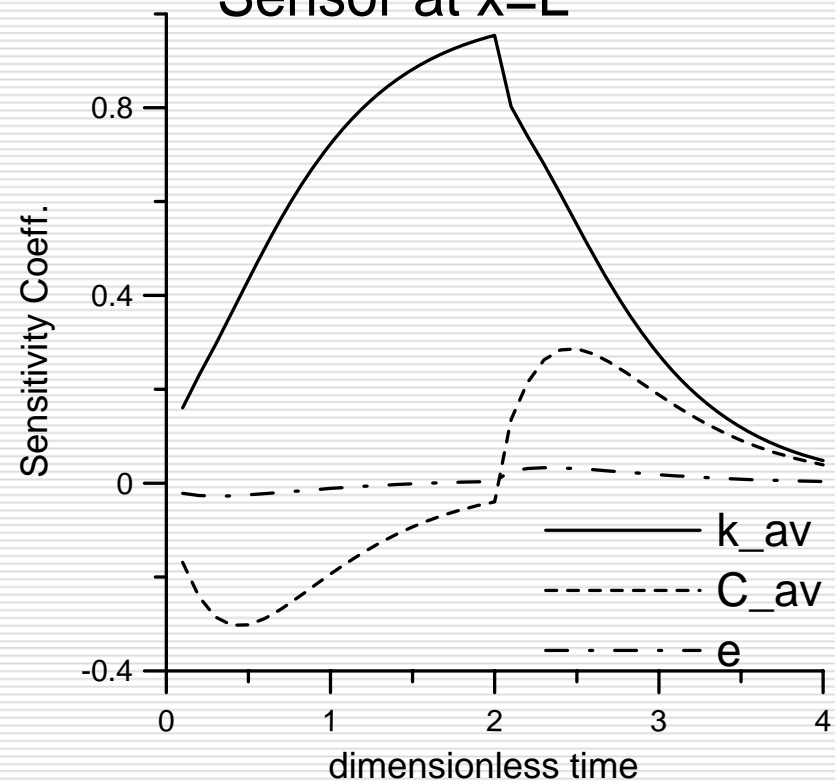
Experiment 1.

Heating at  $x=0$ , two sensors.



Experiment 2.

Sensor at  $x=L$



# Sensitivities and Optimality

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- Sensitivities should be as large as possible.
- Sensitivities must be linearly independent to estimate two or more parameters.
- Given sensitivity matrix  $\mathbf{X}$ , the experiment with the “largest” and “most independent” sensitivities occurs at the maximum value of the determinant of matrix  $\mathbf{X}^T \mathbf{X}$ :

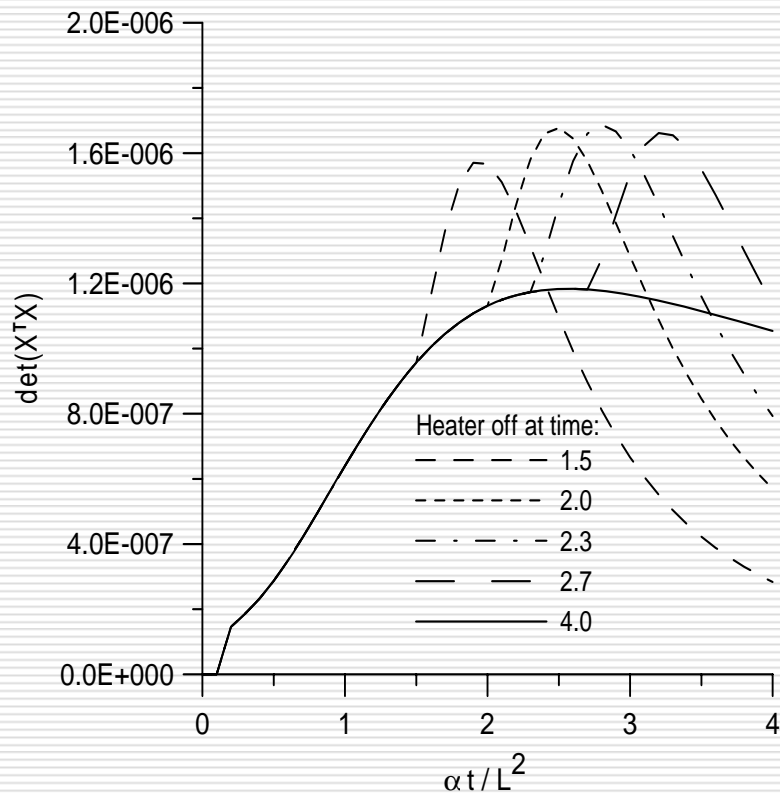
$$D = \frac{1}{s n (T_{max}^+)^2} \det(\mathbf{X}^T \mathbf{X})$$

(s - sensors; n - time steps; )

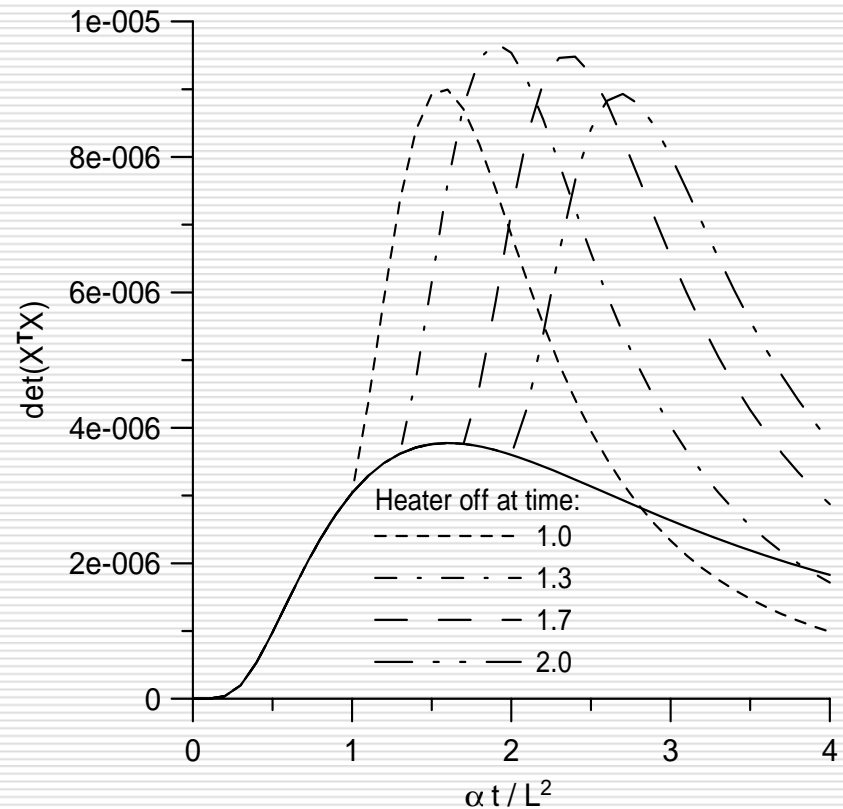


# Optimality Criterion (for $e = 0.2$ ), vary heating duration.

Experiment 1.  
Sensors  $x=0, x = L/4$



Experiment 2.  
Two heating events



# Optimal Experiments.

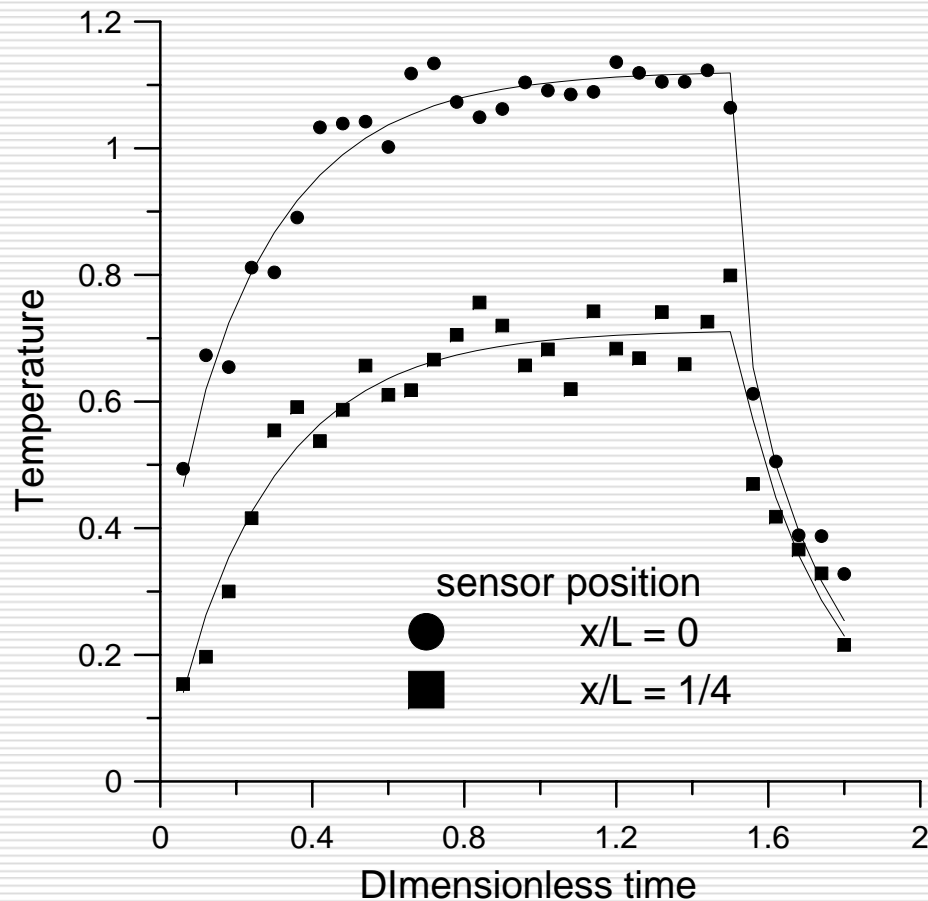
Experiment Design	e	Heat time	Data time	Max. <b>D</b> value
<b>1.</b> Heat $x=0$ , sensors at $x=0$ , $x=L/4$ .	0.2	2.3	2.8	1.7 E-06
	1.0	1.5	1.8	8.7 E-05
	1.8	1.5	1.7	1.2 E-03
<b>2.</b> Two heat events, one sensor each.	0.2	1.3	1.9	9.5 E-06
	1.0	1.7	2.2	24. E-05
	1.8	1.3	1.5	2.2 E-03
<b>3.</b> $k = \text{const.}$	--	2.25	3.0	2.0 E-01

# Simulated data analysis

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- Compute  $T_{\text{exact}}$  vs time from simulation of optimal experiment
- Add noise:  $T_{\text{data}} = T_{\text{exact}} + \varepsilon_i$   
 $\varepsilon_i =$  Zero mean, gaussian, additive error
- Analyze  $T_{\text{data}}$  by a curve fit to the simulation to estimate parameters  $k_{av}$ ,  $C_{av}$ ,  $e$ .

# Simulate Experiment 1, two sensors, error 5% variance



	Exact values	Curve fit
$k_{av}$	1.0	1.0148
$c_{av}$	1.0	1.0080
$e$	1.0	1.0388

# Analysis of optimum experiments (one heat event and two sensors)

slope $e$	Error variance	% error in $k_{av}$	% error in $c_{av}$	% error in $e$
0.2	1%	0.22	0.92	13.9
1.0	1%	0.41	0.79	0.27
1.8	1%	0.39	0.14	0.14
0.2	5%	2.68	7.13	*
1.0	5%	1.48	0.80	3.88
1.8	5%	0.42	3.00	0.33

\* No convergence

# Discuss optimal experiments for FG materials with $k(x)$ linear

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- Best experiment combines two heat events, heat/sensor at  $x=0$  and at  $x=L$ . No interior sensors are needed, non-destructive.
- Length of data record is critical, as optimality criterion  $D=\det(\mathbf{X}^T\mathbf{X})$  has a very narrow peak.
- Optimal heat duration and data record duration vary somewhat with spatial-variation slope,  $e$ .
- Possible to fit three parameters ( $k_{av}$ ,  $c_{av}$ ,  $e$ ) with data from two sensors.
- More accurate parameter estimates are possible for materials with larger variation in  $k$  (larger  $e$ ); it is easier to “see” larger- $e$  values.

# Summary

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- ❑ Design of experiments for finding thermal properties of a functional-graded material with  $k(x)$  across the thickness.
- ❑ Sensitivity to parameters is used to construct optimality criterion  $\det(\mathbf{X}^T\mathbf{X})$  for measurement of thermal properties.
- ❑ Applied to material with linear variation of  $k(x)$ .
- ❑ Presented results of simulated data analysis.

# Future work.

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- Investigate additional materials
  - Other  $k(x)$  distributions (conduction only)
  - Metal foams (conduction and radiation)
- Automate the search for largest value of optimality criterion  $\det(X^T X)$ , to determine the best experimental conditions.
- Is the optimum D-value large enough? Simulate additional cases to estimate parameters from noise-containing data.
- Perform an actual experiment to obtain data on a suitable material; analyze laboratory data.



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