

# An adjoint approach to parameter recovery

## *in a quasilinear parabolic PDE*

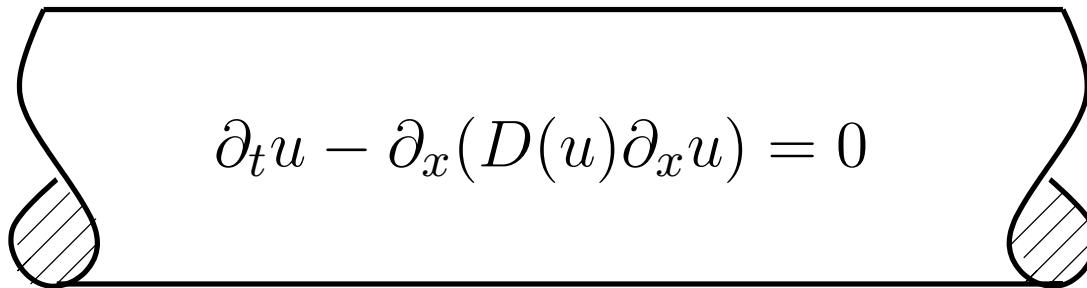
Roger Thelwell (AMATH), Paul DuChateau(CSU), Greg Butters(CSU)

# Outline

- Introduction
- Why?
- Adjoint Theory
- Some numerics
- Conclusions

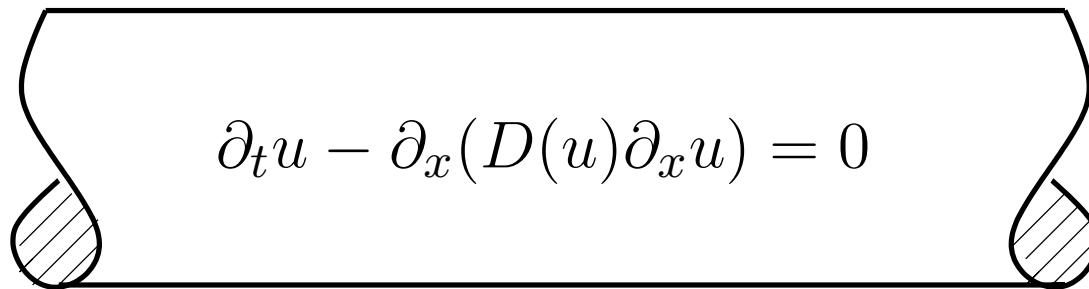
# Introduction

The physical set-up



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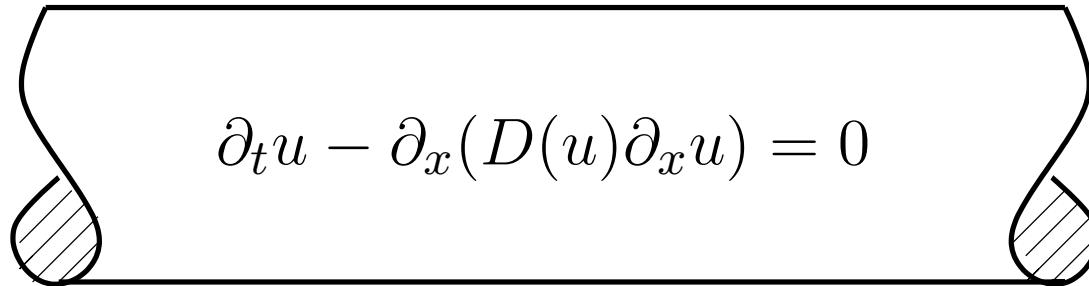


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$$\partial_x u(1, t) = 0$$

# Introduction

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**GOAL:**

Given some output measurements of this system, recover  $D(u)$ .

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- Take  $D(u)$  strictly positive, Lipschitz over  $(0,1)$ , and bounded.
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Let  $W$  represent this class of functions.

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- Why work with toy problem?
- Why consider adjoint methods?

# Forward

The (Forward) Problem: Given  $D(u)$  and  $f(t)$ , find  $u$  which satisfies:

$$\begin{aligned}\partial_t u - \partial_x(D(u)\partial_x u) &= 0 & (x, t) \in U_T \\ u(0, t) &= f(t) & t \in (0, T) \\ \partial_x u(1, t) &= 0 & t \in (0, T) \\ u(x, 0) &= f(0) & x \in (0, 1)\end{aligned}$$

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Why?

- Unique weak (forward) solution if  $D$  and  $f$  from  $W$ .
- Model of a simple physical process.
- Measurements of flux at  $x = 0$  and state at  $x = 1$  possible.

# Inverse

The (Inverse) Problem: Given boundary measurements  $g(t)$  and  $h(t)$ , recover  $D(u)$  which satisfies:

$$\begin{aligned}\partial_t u - \partial_x(D(u)\partial_x u) &= 0 & (x, t) \in U_T \\ -D(u)\partial_x u(0, t) &= g(t) & t \in (0, T) \\ u(1, t) &= h(t) & t \in (0, T) \\ u(x, 0) &= f(0) & x \in (0, 1)\end{aligned}$$

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Why?

- Experimental interest.
- Simple measurements.
- If we can recover a  $D$  from  $W$ , it should be unique.

# Adjoint

## Max Min Statement

- for each  $t$ ,  $f(0) < u(x, t) < f(t)$ ,  $0 < x < 1$
- $\partial_x u(x, t) < 0$  a.e. on  $(0, 1) \times (0, T)$

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Comments:

If we let  $B(u) = \int_0^u D(s) ds$ , then  $D(u)\partial_x u = \partial_x B(u)$ .

Also,  $B(u) = \int_0^u D(s) ds = (u - 0) D(\tilde{u})$  for some  $\tilde{u}$  between  $u$  and 0.

# Adjoint

Again, let  $g(t) := -D(u(0, t))\partial_x u(0, t)$  and  $\Phi[D, f] = g(t)$ .

We can show that if  $D_1(u) > D_2(u)$  then the map  $\Phi$  is:

- Monotone :  $\Phi[D_1, f](t) > \Phi[D_2, f](t)$

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- Continuous:  $\|g_1 - g_2\|_{L^2(0,T)} \leq C\|D_1 - D_2\|_\infty$
- Injective :  $\Phi[D_1, f] = \Phi[D_2, f]$  implies  $D_1 = D_2$

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# Adjoint

Integral Identity:

Let  $u, v$  represent solutions to direct problem with  $D = D_1, D_2$ , respectively. Subtract, multiply by smooth  $\phi$  and integrate by parts to yield:

$$\begin{aligned} & \iint_{U_T} (u - v)(\phi_t + k\phi_{xx}) - \int_0^1 (u - v)\phi \Big|_{t=0}^{t=T} + \\ & + \int_0^T (D_1(u)u_x - D_2(v)v_x)\phi \Big|_{x=0}^{x=1} + \int_0^T (D_1(u) - D_1(v))(k\phi_x) \Big|_{x=0}^{x=1} \\ & = \iint_{U_T} (D_1(v) - D_2(v)v_x)\phi_x \end{aligned}$$

Here  $U_T = (0, 1) \times (0, T)$  and  $k := D(\tilde{u})$  for some  $\tilde{u}$  in the interval  $(v, u)$

# Adjoint

The general adjoint problem:

$$\begin{aligned}\partial_t \phi + k \partial_{xx} \phi &= F^*(x, t) & (x, t) \in U_T \\ \phi(x, T) &= p^*(x) & x \in (0, 1) \\ k \phi_x &= g^*(t) & x = 0, t \in (0, T). \\ \phi &= h^*(t) & x = 1, t \in (0, T)\end{aligned}$$

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# Adjoint

The dual data acts like switches. Data appears in a duality pairing with terms of the direct problem.

Taking  $F^* = 0, p^* = 0, h^* = 0$ , the identity collapses to

$$\int_0^T g^*(g_1 - g_2) = \iint_{U_T} (D_1(v) - D_2(v)) v_x \phi_x$$

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Letting  $\Phi[D, f] = g(t)$ , we could cast this formally as:

$$\begin{aligned} (\Phi[D_1, f] - \Phi[D_2, f], \theta)_{L^2} &:= (\delta\Phi[D_1, D_2]\Delta D, \theta)_{L^2} \\ &= \langle \Delta D, {}^t\delta\Phi[D_1, D_2] \theta \rangle_{W \times W^*} \end{aligned}$$

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The integral identity provides a realization of this expression.

$$\int_0^T g^*(g_1 - g_2) = \iint_{U_T} (D_1(v) - D_2(v)) v_x \phi_x$$

# Adjoint

Now some quick proofs:

$$\int_0^T g^*(g_1 - g_2) dt = \iint_{U_T} (D_1(v) - D_2(v)) v_x \phi_x dx dt$$

- Monotone: Assume  $D_1 > D_2$ . We know  $v_x < 0$ . Taking  $g^* > 0$  implies  $\phi_x < 0$ . So  $g_1(t) > g_2(t)$ .

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- **Continuous:** Take  $g^* = (g_1 - g_2)/||g_1 - g_2||_{L^2}$ .
- **Injective:** Assume  $D_1 - D_2 \not\equiv 0$ . This implies that  $g_1 - g_2 \not\equiv 0$ .

# Numerics

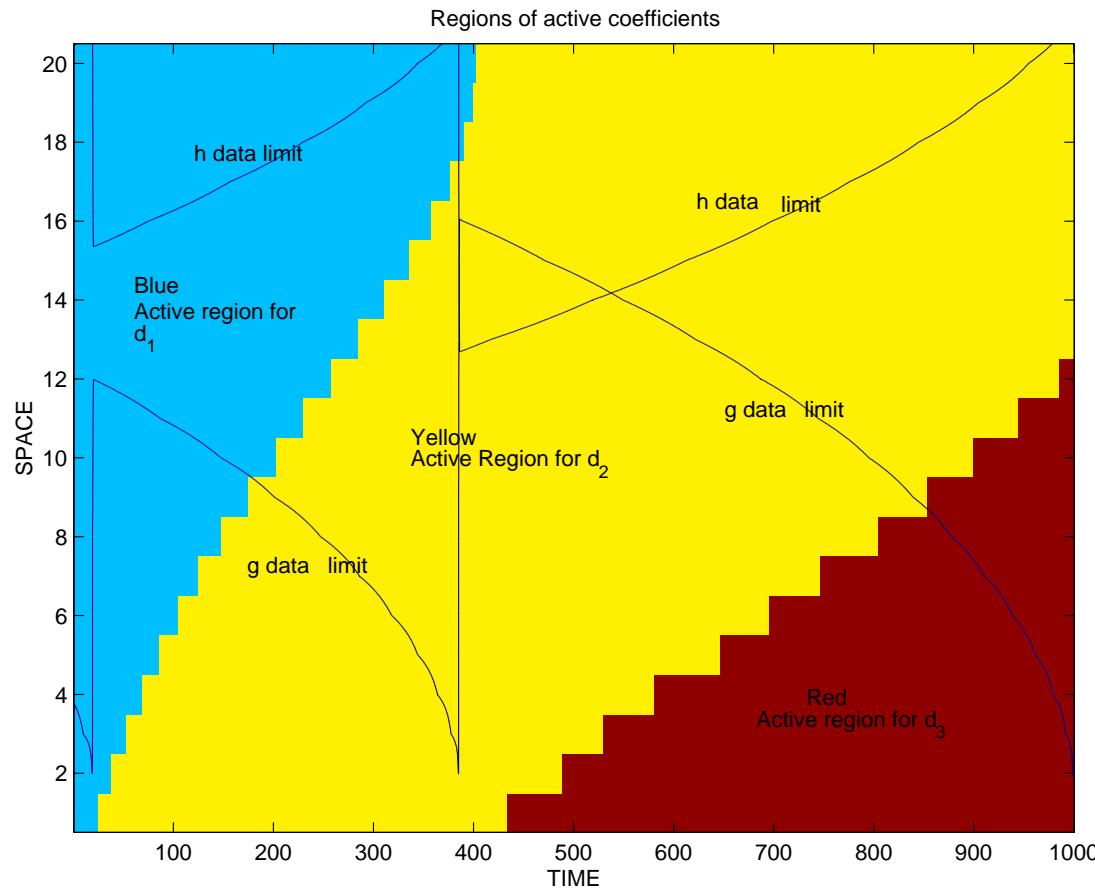
If  $f(t)$  is strictly monotone in time, a time discretization induces a discretization on  $f(t)$ . A weak max principle then allows  $u$  to be discretized. This leads to parameterization for D:

$$\hat{D}(u) = \sum_{k=0}^N d_k \Lambda_k,$$

where  $\Lambda_k$  given by the standard hat function. The  $\hat{D}(u)$  direct problem is uniquely solvable.

# Numerics

## Regions of coefficient activity



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- Given  $g^*(t)$  and  $k$ , solve adjoint problem to generate  $\phi(x, t)$

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- Given  $g^*(t)$  and  $k$ , solve adjoint problem to generate  $\phi(x, t)$
- Apply integral ID to compute update for  $D$ .
- Repeat

# Numerics

Apply identity to region where only  $d_1$  is active.

$$b_1 = \int_0^{t_1} g^* \Delta g = (d_1 - \delta_1) \iint_{R_{11}} \Lambda_1(u) u_x \phi_x = A_{11} \Delta d_1$$

where  $R_{11} = \{(x, t) \in U_{t_1} : u_0 < u(x, t) < u_1\}$

# Numerics

Apply identity to the  $U_{t_2}$  time strip. Only  $d_1$  and  $d_2$  are active.

$$\begin{aligned} b_2 := \int_{t_1}^{t_2} g^* \Delta g &= (d_2 - \delta_2) \iint_{R_{22}} \Lambda_2(u) u_x \phi_x + \\ &\quad + (d_1 - \delta_1) \iint_{R_{21}} \Lambda_1(u) u_x \phi_x \\ &= A_{22} \Delta d_2 + A_{21} \Delta d_1 \end{aligned}$$

with

$$\begin{aligned} R_{21} &= \{(x, t) \in U_{t_2} : u_0 < u < u_1\} \\ R_{22} &= \{(x, t) \in U_{t_2} : u_1 < u < u_2\} \end{aligned}$$

# Numerics

By the  $Q_{t_N}$  strip, a diagonal system forms.

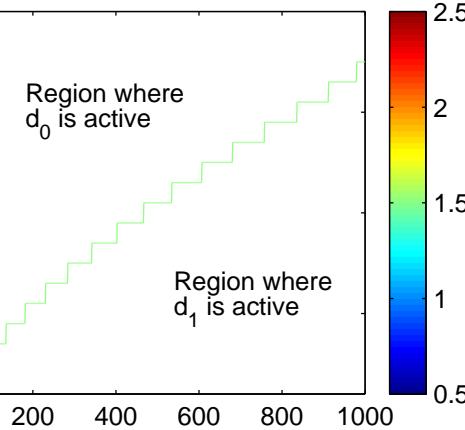
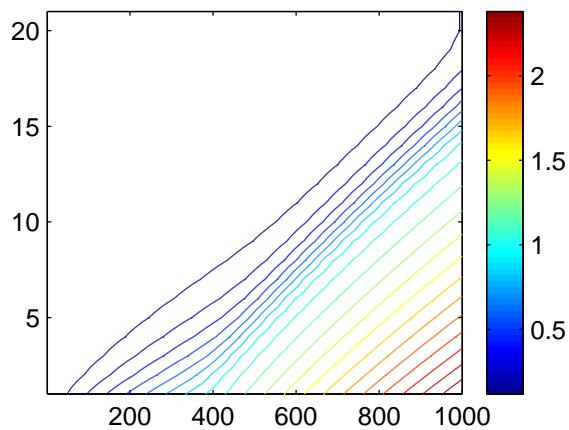
$$\begin{bmatrix} A_{11} & 0 & \cdots & 0 \\ A_{21} & A_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & 0 \\ A_{N1} & \cdots & \cdots & A_{NN} \end{bmatrix} \begin{bmatrix} \Delta d_1 \\ \Delta d_2 \\ \vdots \\ \Delta d_N \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_N \end{bmatrix}$$

where  $A_{ij} = \iint_{R_{ij}} \Lambda_j(u) u_x \phi_x$  and  $b_i = \int_{t_{i-1}}^{t_i} g^* \Delta g$

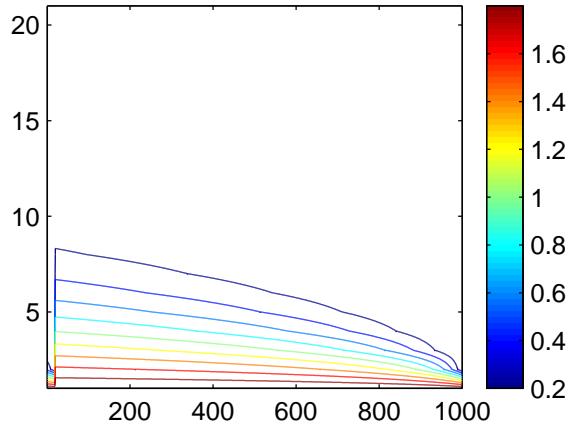
# Numerics

## Coefficients ranges

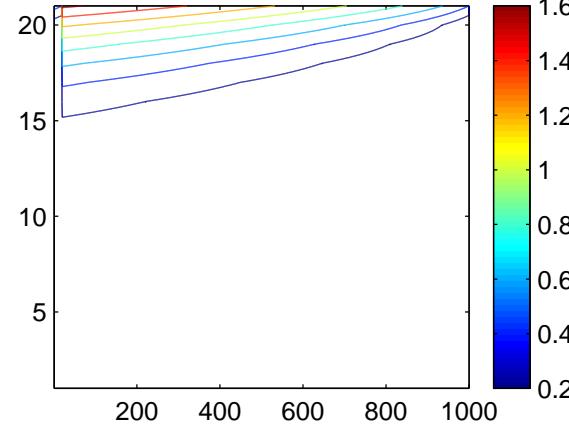
Direct with  $u(0,t)=F$  &  $u_x(1,t)=0$



Dual soln with h BCs



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# Numerics

Demos: is some data better? Use  $g^*$  data,  $h^*$  data or some combination?

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- I'm happy to be done. Thanks!