

p -Laplacian solved, and line search
it's nice for SNES to have an objective

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outline for today

Chapter 5 of book:

- ▶ recall p -Laplacian equation

$$-\nabla \cdot (|\nabla u|^{p-2} \nabla u) = f$$

- arises from minimizing objective $I[u] = \int_{\Omega} \frac{1}{p} |\nabla u|^p - fu$
- discretized with Q^1 structured-grid finite element method
- ▶ **code** `c/ch5/plap.c`:
 - objective-only prototype implementation:
 - ▶ i.e. option `-snes_fd_function`
 - ▶ severe scalability problems
 - add residual to succeed:
 - ▶ compare `-snes_fd` and `-snes_fd_color`
 - ▶ show clear evidence of convergence
 - ▶ compare quadrature degree
- ▶ line search for Newton's method

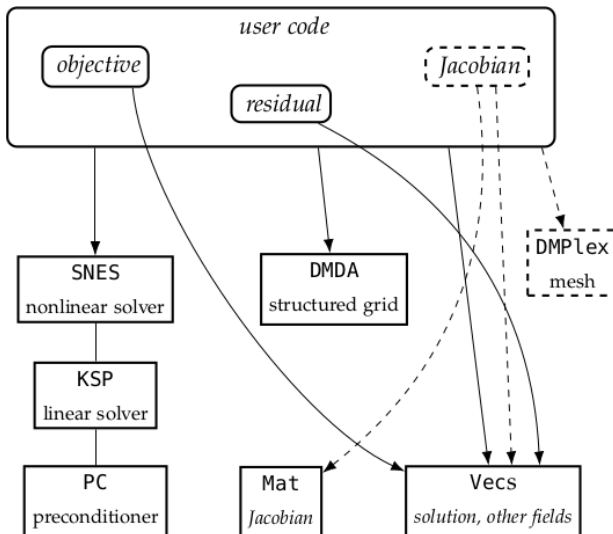
optimization vs solving equations

- ▶ it is possible to lose track of which “level” stuff is at
- ▶ a correspondence (N = number of unknowns):

output dim	<i>optimization</i>	\leftrightarrow	<i>solving equations</i>
1	$\min I[u]$ <u>objective</u>		?
N	$\nabla I[u] = 0$ <u>gradient</u>		$\mathbf{F}(u) = 0$ <u>residual</u>
N^2	$H_{ij}(u) = \frac{\partial^2 I}{\partial u_i \partial u_j}(u)$ <u>Hessian</u>		$J_{ij}(u) = \frac{\partial F_i}{\partial u_j}(u)$ <u>Jacobian</u>

- ▶ `c/ch5/plap.c` implements
 - `objective` in `FormObjectiveLocal()`
 - `gradient (=residual)` in `FormFunctionLocal()`
 - but *not* `Hessian (=Jacobian)`

structure of c/ch5/plap.c



with objective-only: breaks easily

- ▶ try it out:

```
$ cd c/ch5
$ make plap
$ ./plap -snes_monitor -snes_fd_function -snes_fd
$ ./plap -snes_monitor -snes_fd_function -snes_fd_color
```

- ▶ flaky behavior under refinement; if LEV= 1, 2, 3 then

```
$ ./plap -snes_monitor -snes_fd_function -snes_fd_color \
    -snes_converged_reason -da_refine $LEV
```

gives DIVERGED for 1,3 levels and CONVERGED for 2 level

- ▶ fixable with weaker Newton tolerance:

```
$ ./plap -snes_monitor -snes_fd_function -snes_fd_color \
    -snes_converged_reason -da_refine $LEV \
    -snes_linesearch_type basic -snes_rtol 1e-6
```

- ▶ ...but this is unsustainable (next slide)

with objective-only: too many objective evaluations

- ▶ count evaluations of `FormObjectiveLocal()` = $I^h[u]$

- ▶ for `LEV= 0, 1, 2, 3` do:

```
$ ./plap -snes_fd_function -snes_fd_color \  
    -snes_linesearch_type basic -snes_rtol 1e-6 \  
    -log_view -da_refine $LEV |grep Eval
```

- ▶ result:

<i>level</i>	<i>grid</i>	<i>error</i>	<i># of objective evals</i>
0	3×3	8.1×10^{-3}	1428
1	5×5	2.9×10^{-3}	4636
2	9×9	9.3×10^{-4}	12444
2	17×17	2.6×10^{-4}	104160

- ▶ evidence for convergence at $O(\Delta x^2)$ or $O(\Delta x^{1.5})$??
- ▶ but it is too much work per grid point

back to calculus: get gradient of objective

- ▶ we can differentiate the thing we are optimizing:

$$\begin{aligned} I[u + \epsilon v] - I[u] &= \int_{\Omega} \frac{1}{p} |\nabla u + \epsilon \nabla v|^p - \frac{1}{p} |\nabla u|^p - \epsilon f v \\ &= \epsilon \left(\int_{\Omega} |\nabla u|^{p-2} \nabla u \cdot \nabla v - f v \right) + O(\epsilon^2). \end{aligned}$$

- if u is minimum of $I[u]$ then $I[u + \epsilon v] - I[u] \geq 0$

- ▶ get formula for the *gradient*:

$$\nabla I[u](v) = \lim_{\epsilon \rightarrow 0} \frac{I[u + \epsilon v] - I[u]}{\epsilon} = \int_{\Omega} |\nabla u|^{p-2} \nabla u \cdot \nabla v - f v$$

- for each $u \in W^{1,p}(\Omega)$ it's a map $\nabla I[u] : W^{1,p}(\Omega) \rightarrow \mathbb{R}$

p -Laplacian weak formulation: $\mathbf{F}(u) = 0$

- ▶ the *equations* we want to solve are “derivative = 0”:

$$\nabla I[u](v) = 0 \quad \forall v \quad \Longleftrightarrow \quad \mathbf{F}(u) = 0$$

- ▶ equation is the *weak form*

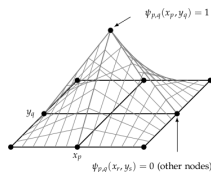
$$\int_{\Omega} |\nabla u|^{p-2} \nabla u \cdot \nabla v - f v = 0$$

- each $v \in W_0^{1,p}(\Omega)$ gives an equation which unknown $u \in W_g^{1,p}(\Omega)$ must satisfy
 - $F_i(u) = \int_{\Omega} |\nabla u|^{p-2} \nabla u \cdot \nabla v_i - f v_i$ is a scalar component of the residual
- ▶ additional integration-by-parts gives *strong form*

$$-\nabla \cdot (|\nabla u|^{p-2} \nabla u) = f$$

discrete weak form is residual

- ▶ hat functions $\psi_{pq}(x, y)$ form basis of S_0^h



- ▶ we want $u \in S_g^h$ so that $\nabla I^h[u](\psi_{pq}) = 0$ for all p, q
- ▶ integral over Ω is sum over elements \square_{ij} :

$$\sum_{i,j} \int_{\square_{ij}} |\nabla u|^{p-2} \nabla u \cdot \nabla \psi_{pq} - f \psi_{pq} = 0$$

- ▶ each grid location p, q gives one component of residual:

$$F_{pq}(u) = \sum_{i,j} \int_{\square_{ij}} |\nabla u|^{p-2} \nabla u \cdot \nabla \psi_{pq} - f \psi_{pq}$$

residual on one element

- ▶ need residual on element \square_{ij}

$$H_{ij}^{pq}(x, y) = \left[|\nabla u|^{p-2} \nabla u \cdot \nabla \psi_{pq} - f \psi_{pq} \right]_{\square_{ij}}$$

- ▶ ... but we want it on the ref. element, for quadrature
- ▶ only four ψ_{pq} are nonzero on \square_{ij}
 - correspond to element corners $\ell = 0, 1, 2, 3$:

$$\psi_{pq} = \chi_{\ell}$$

- so define:

$$H_{ij}^{\ell}(\xi, \eta) = \left[|\nabla u|^{p-2} \nabla u \cdot \nabla \chi_{\ell} - f \chi_{\ell} \right]_{\square_*}$$

- details in Exercise 5.8

discrete weak form with quadrature

- ▶ recall Gauss-Legendre quadrature

$$\begin{aligned}\int_{\square_{ij}} v(x, y) \, dx \, dy &= \frac{h_x h_y}{4} \int_{\square_*} v(\xi, \eta) \, d\xi \, d\eta \\ &\approx \frac{h_x h_y}{4} \sum_{r=0}^{n-1} \sum_{s=0}^{n-1} w_r w_s v(z_r, z_s)\end{aligned}$$

- ▶ putting it all together, we loop over elements i, j and add each element's contribution to the p, q residual:

$$F_{p,q}(u) += \frac{h_x h_y}{4} \sum_{r=0}^{n-1} \sum_{s=0}^{n-1} w_r w_s H_{ij}^\ell(z_r, z_s),$$

- if the ℓ corner of \square_{ij} is node p, q
 - details in `FormFunctionLocal()`
- ▶ then we solve $N = m_x m_y$ equations

$$\mathbf{F}(u) = 0$$

it makes all the difference

- ▶ we can count evaluations of `FormObjectiveLocal()` = $I^h[u]$ and `FormFunctionLocal()` = $\mathbf{F}(u)$
- ▶ objective-only:

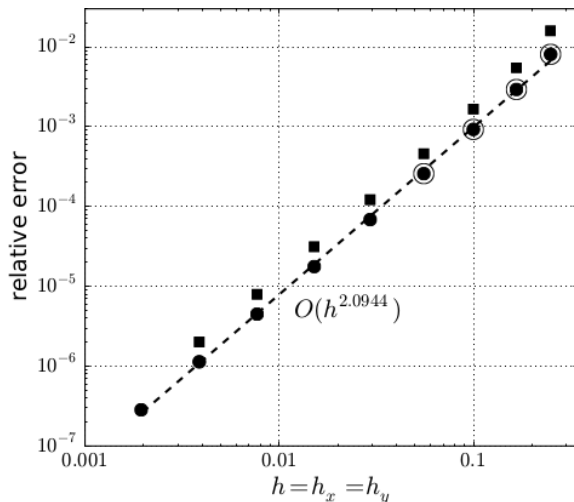
```
$ ./plap -da_refine 1 -log_view \  
      -snes_fd_function -snes_fd_color |grep Eval
```

gives 5333 objective evals
- ▶ with residual \mathbf{F} :

```
$ ./plap -da_refine 1 -log_view |grep Eval
```

gives 8 objective (why?) and 45 function evals
- ▶ two versions give same numerical error
- ▶ residual version gives fewer Newton steps (why?)

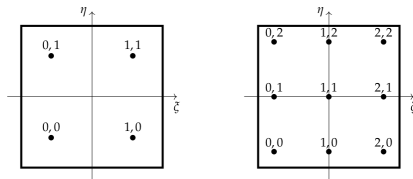
convergence



- ▶ dots for residual version, circles for objective-only
- ▶ convergence rate: good!

effect of quadrature

- recall we implemented $n = 1, 2, 3$ order Gauss-Legendre quadrature in `plap.c`



- compare numerical error:

```
$ timer ./plap -da_refine 4 -plap_quaddegree 1  
numerical error: |u-u_exact|/|u_exact| = 1.219e-04  
real 0.20
```

```
$ timer ./plap -da_refine 4 -plap_quaddegree 2  
numerical error: |u-u_exact|/|u_exact| = 6.863e-05  
real 0.44
```

```
$ timer ./plap -da_refine 4 -plap_quaddegree 3  
numerical error: |u-u_exact|/|u_exact| = 6.863e-05  
real 0.86
```

- squares on prev graph show $n = 1$ quadrature results
- conclude: $n = 2$ good ... thus it is the default

line search

- ▶ recall Newton's method:

$$J_F(\mathbf{x}_k) \mathbf{s} = -\mathbf{F}(\mathbf{x}_k)$$

$$\mathbf{x}_{k+1} = \mathbf{x}_k + \mathbf{s}$$

- ▶ full step \mathbf{s}_k may not reduce $I[\mathbf{x}]$... if there is an objective function $I[\mathbf{x}]$ at all!
- ▶ options:
 - if there is objective function $I[\mathbf{x}]$ then do line search

$$\min_{\lambda > 0} I[\mathbf{x}_k + \lambda \mathbf{s}_k]$$

- if not, define *merit function* $\phi(\mathbf{x}) = \frac{1}{2} \|\mathbf{F}(\mathbf{x})\|_2^2$ and then do line search

$$\min_{\lambda > 0} \phi(\mathbf{x}_k + \lambda \mathbf{s}_k)$$

line search control in PETSc

- ▶ options for `-snes_linesearch_type`:

<u>Name</u>	<u>Summary</u>
<code>basic</code>	no line search; use full Newton step $\lambda_k = 1$
<code>bt</code> [<i>default</i>]	polynomial-fit back-tracking; uses obj. if present
<code>cp</code>	assume F is gradient of obj. and find crit. point
<code>l2</code>	secant-line minimization; fixed repeats

- ▶ use `-snes_linesearch_monitor` to see action