# Numerical methods

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Consider the govering equations

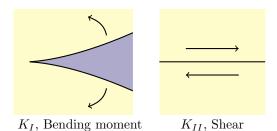
$$\begin{pmatrix} p(z) \\ 0 \end{pmatrix} = \begin{pmatrix} \sigma_y \\ \tau_{xy} \end{pmatrix} = \int_0^\infty \begin{pmatrix} K_{11}(x-z) & K_{12}(x-z) \\ K_{21}(x-z) & K_{22}(x-z) \end{pmatrix} \begin{pmatrix} g'(x) \\ h'(x) \end{pmatrix} dx \quad (1)$$

$$h^2 p' = \lambda \tag{2}$$

Have the "input" parameters as

- BC's P, M (or equivalently g', h'' at  $x \to \infty$ )
- $\lambda$  , the speed

Want to solve for the toughness  $K_I$  and  $K_{II}$ . In this project, we have so far focused on  $K_I$ .



<u>Goal</u>: Find  $\lambda$  such that  $K_I(\lambda) = 0$ , "Zero toughness solution". Given this we then want to investigate the behaviour for small  $K_I \approx 0$ . To do this, take some given value of  $\lambda$  and then solve equations 1, 2.

#### Discretization of problem

The method chosen to discretize the problem is to take a vector  $(x_1, \ldots, x_n)$  of n points at which we measure g', h' and have a vector  $(z_1, \ldots, z_{n-1})$  of n-1 intermediate points at which p is measured. (The spacing chosen is a  $\tan^2$  spacing)

The "obvious" way to interpolate h' in between the  $x_i$ 's is simple linear interpolation. But both h', g' become singular near 0. However, we expect a  $x^{-1/2}$ 

x, the n points at which h', g' are measured



z, the n-1 points at which p is measured

singularity, which allows us to "remove" said singularity. The interpolation used is

$$g'(x) = \begin{cases} \frac{1}{\sqrt{x}} (a_i x + b_i) & i < t \\ a_i x + b_i & i \ge t \end{cases}$$

for x in the spline  $x \in [x_i, x_{i+1}]$ . Choose 1 < t < n, typically t = n/2. Similarly

$$h'(x) = \begin{cases} \frac{1}{\sqrt{x}} (c_i x + d_i) & i < t \\ c_i x + d_i & i \ge t \end{cases}$$

With the same t used. We also define  $a_n, b_n, c_n, d_n$  for interpolation beyond  $x_n$ . The values of g', h' are stored via

$$\boldsymbol{\theta} = \begin{pmatrix} a_1 x_1 + b_1 \\ \vdots \\ a_n x_n + b_n \\ c_1 x_1 + d_1 \\ \vdots \\ c_n x_n + d_n \end{pmatrix}$$

Once one has  $\boldsymbol{\theta}$ , it is trivial to recover, say  $g'(x_i)$ , since either  $g'(x_i) = \boldsymbol{\theta}_i$  or  $g'(x_i) = \boldsymbol{\theta}_i / \sqrt{x_i}$ . Similarly, given  $g'(x_i)$ ;  $\boldsymbol{\theta}_i$  can be calculated.

#### Recovering the $a_i$ 's

Suppose we know  $\theta$ , (and always assume we know the  $x_i$ ). Can we recover  $a_i, b_i, c_i, d_i$ ? The answer is yes, once we add in the boundary conditions at  $\infty$ .

Further we have that

$$oldsymbol{\gamma} = \left( egin{array}{c} a_1 \ dots \ a_n \ b_1 \ dots \ b_n \ c_1 \ dots \ c_n \ d_1 \ dots \ d_n \ \end{array} 
ight) = Toldsymbol{ heta}$$
 interpolation matrix. A quick

Where T is a  $4n \times n$  interpolation matrix. A quick check reveals we have 4n unknowns, in  $\gamma$ . Knowing  $\theta$  provides 2n equations. Demanding continuity of the interpolated g', h' provides another 2(n-1) equations, (match at  $x_2, \ldots, x_n$ ). Finally boundary conditions on the spline at  $\infty$  provide another 2 equations.

The continuity conditions are

$$a_{1}x_{2} + b_{1} = a_{2}x_{2} + b_{2}$$

$$\vdots$$

$$a_{t-2}x_{t-1} + b_{t-2} = a_{t-1}x_{t-1} + b_{t-1}$$

$$(a_{t-1}x_{t} + b_{t-1})/\sqrt{x_{t}} = a_{t}x_{t} + b_{t}$$

$$a_{t}x_{t+1} + b_{t} = a_{t+1}x_{t+1} + b_{t+1}$$

$$\vdots$$

$$a_{n-1}x_{n} + b_{n-1} = a_{n}x_{n} + b_{n}$$

and similar for c, d. This means that with the exceptions of i = t - 1, n, have that

$$\frac{\theta_{i+1}-\theta_i}{x_{i+1}-x_i}=a_i$$
 
$$\frac{\theta_i\,x_{i+1}-\theta_{i+1}x_i}{x_{i+1}-x_i}=b_i$$

Same idea with  $x_{t-1}$ , just have to be a little careful about the switch in the continuity condition,

$$\frac{\sqrt{x_t}\theta_t - \theta_{t-1}}{x_t - x_{t-1}} = a_{t-1}$$
 
$$\frac{\theta_{t-1} x_t - \theta_t x_{t-1} \sqrt{x_t}}{x_t - x_{t-1}} = b_{t-1}$$

So we are almost done, just missing 4 rows in our matrix. Have the  $n^{th}$  row as all zeros, i.e.  $a_n = 0$  due to boundary conditions, and so trivially the  $2n^{th}$  row is 0 except  $T_{2n,n} = 1$  Now, B.C. for h' implies  $h''(x_n) \approx h''(x_{n-1})$  i.e.  $c_n = c_{n-1}$  and thus from continuity  $d_n = d_{n-1}$ . This completes our interpolation matrix T. We still have some extra boundary conditions to impose, which we will do later. Naively, these are  $g'(x_n) = 1/2$ ,  $c_n = 1$ . These are approximately true, but we need to do a bit better. (I will add this into the explanation once I understand it...)

#### Analytic expressions

Now that we have made the piecewise analytic approximation<sup>1</sup> we can avoid making any more approximations. Recall  $K_{ij}$  has an analytic expression (even better, it's a rational function). If we take h', g' to be piecewise analytic, then the integrand becomes an analytic expression that can be exactly integrated. For example,

$$p(z) = \int_0^\infty K_{11}(x-z)g'(x) + K_{12}(x-z)h'(x) dx$$

$$= \sum_{i=1}^{t-1} \int_0^\infty K_{11}(x-z)(a_ix+b_i)/\sqrt{x} + K_{12}(x-z)(c_ix+d_i)/\sqrt{x} dx$$

$$+ \sum_{i=t}^n \int_0^\infty K_{11}(x-z)(a_ix+b_i) + K_{12}(x-z)(c_ix+d_i) dx$$

The right hand side of this equation may look ghastly, (and we haven't even expanded the  $K_{ij}$ 's yet ...) but it is an analytic expression in z. Further, we can do the integration before knowing any of the values of a, b, c, d. It is also clear that p(z) is linear in a, b, c, d. Therefore, once we know the spacing of the  $z_i$ 's we see that

$$\begin{pmatrix} p(z_1) \\ \vdots \\ p(z_{n-1}) \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} B_{1,1} & \cdots & B_{1,2n} \\ \vdots & \ddots & \vdots \\ B_{2(n-1),1} & \cdots & B_{2(n-1),2n} \end{pmatrix} \gamma = BT\boldsymbol{\theta}$$

Where the matrix B depends on the choice of spacings (x, z), but does not depend on  $\gamma$ .

 $<sup>^{-1}</sup>$ I think I have made up some terminology here, piecewise linear isn't quite right due to the  $x^{-1/2}$  parts, and "sometimes piecewise linear sometimes piecewise  $x^{-1/2} \times (\text{linear function})$ " is not ideal either

We can go further and incorporate the boundary conditions into this equation. The discretized versions of the boundary conditions, become  $g'(x_n) = 1/2$  and  $\frac{h'(x_n) - h'(x_{n-1})}{x_n - x_{n-1}} = 1$ . These conditions are linear in terms of g', h', and so by adding another two rows onto the matrix BT, get that

$$\begin{pmatrix} p(z_1) \\ \vdots \\ p(z_{n-1}) \\ 0 \\ \vdots \\ 0 \\ g'(\infty) \\ h''(\infty) \end{pmatrix} = \begin{pmatrix} A_{1,1} & \cdots & A_{1,2n} \\ \vdots & \ddots & \vdots \\ A_{2n,1} & \cdots & A_{2n,2n} \end{pmatrix} \boldsymbol{\theta}$$

Where  $g'(\infty), h''(\infty)$  are the (constant) boundary conditions.

Now we use the second equation for p, namely  $p = \int_z^\infty \lambda/h^2 dx$ . We can integrate our piecewise analytic expression for h' to recover h, imposing both  $h(x_1) = 0$  as well as continuity at the  $x_i$  for i = 2, ... n. The result is

$$h(x) = \begin{cases} \sqrt{x}(w_i x + e_i) + r_i & i < t \\ w_i x^2 + e_i x + r_i & i \ge t \end{cases}$$

for some constants w, e, r. These constants are related to  $\gamma$ , and so  $\theta$ , linearly. Given this piecewise analytic expression for h, we can find an analytic expression for p(z). Since we know the spacings, we have

$$f(\boldsymbol{\theta}) = \begin{pmatrix} p(z_1) \\ \vdots \\ p(z_{n-1}) \\ 0 \\ \vdots \\ 0 \\ g'(\infty) \\ h''(\infty) \end{pmatrix} = A\boldsymbol{\theta}$$

Where we now just need to solve for  $\theta$ .

#### Newton's method

Suppose  $\theta$  is iterate 1. To get the next iterate you need to solve (to first order)

$$f(\boldsymbol{\theta} + \delta \boldsymbol{\theta}) = A(\boldsymbol{\theta} + \delta \boldsymbol{\theta})$$

 $<sup>^2</sup>$ Again, we need to do a bit better than this. For now, this illustrates the point.

$$f(\boldsymbol{\theta}) + (Df|_{\boldsymbol{\theta}})(\delta\boldsymbol{\theta}) = A\boldsymbol{\theta} + A\delta\boldsymbol{\theta}$$

Where  $Df|_{\theta}$  is a matrix of partial derivatives. Therefore, get to first order that

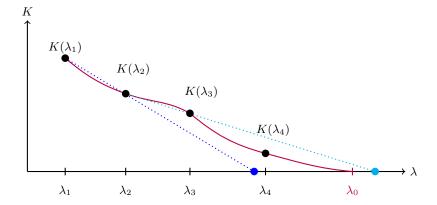
$$\delta \boldsymbol{\theta} = (A - Df|_{\boldsymbol{\theta}})^{-1} (f(\boldsymbol{\theta}) - A\boldsymbol{\theta})$$

Ingredients:

- Matrix A itself (of which the  $2(n-1) \times 2n$  part is the integral kernel)
- The function  $f(\theta)$ . I.e. given  $\theta$  you need to calculate  $\int_z^\infty \lambda/h^2 dx$  (Key functions "hprime\_to\_h" and "hprime\_to\_p").
- Need to calculate Df which involves calculating  $\frac{\partial}{\partial \theta} \int_{z}^{\infty} \frac{\lambda}{h(x)^2} dx$

So we have worked out numerically  $K(\lambda)$ , now we want to solve  $K(\lambda_0) = 0$  for  $\lambda_0$ . We do a "march". Sublety in that K < 0 is unphysical, so a guess of  $\lambda > \lambda_0$  where  $K(\lambda_0) = 0$  does not make any physical sense (& will get bad numerical results). To get around this difficulty, take the next iterate of  $\lambda$  as smaller than predicted.

Figure 1: March to find  $\lambda_0$ 



For example in Figure 1, the obvious choice for  $\lambda_4$  (the light blue circle) is larger than the true value of  $\lambda_0$ , and therefore the naive extrapolation method won't quite work.

## Guide to programs

#### K\_of\_c\_march

First the program sets up the spacing as  $\tan^2$ . It also sets the initial  $h' = \underbrace{(1,\ldots,1}_{g'}\underbrace{x_1+1,\ldots,x_n+1}_{h'})$  (Not sure why this is a reasonable first guess, per-

haps from the boundary conditions at  $\infty$ . Seems to have no problems converging though.)

Most of the work is then done by fixed\_lambda\_M\_iteration which then solves for  $K_I$  and h'.

N.B.  $\boldsymbol{h}'$  is updated via  $\boldsymbol{h}'_i = \frac{\boldsymbol{h}'_{i-1} - \boldsymbol{h}'_{i-2}}{\lambda_{i-1} - \lambda_{i-2}} \lambda_i + \frac{\lambda_{i-1} \boldsymbol{h}'_{i-2} - \lambda_{i-1} \boldsymbol{h}'_{i-1}}{\lambda_{i-1} - \lambda_{i-2}}$  which is just linear extrapolation. In the absence of any better ideas this is the sensible choice.

After iterating for a few values, get near  $\lambda_0$ . Hear we suspect that something like  $K^3 \sim \lambda - \lambda_0$  near  $\lambda = \lambda_0$ , K = 0. So given two prior guesses, extrapolate via  $\lambda_i = \frac{K_{i-1}^3 \lambda_{i-2} - K_{i-2}^3 \lambda_{i-1}}{K_{i-1}^3 - K_{i-2}^3}$  But as noted earlier, must be careful to not extrapolate further than  $\lambda_0$ . So an idea is to take  $(\lambda_i + \lambda_{i-1})/2$  as the next guess, i.e.

$$\lambda_i = \frac{\lambda_{i-1} - \lambda_{i-2}}{K_{i-1}^3 - K_{i-2}^3} \frac{K_{i-1}^3}{2} + \frac{K_{i-1}^3 \lambda_{i-2} - K_{i-2}^3 \lambda_{i-1}}{K_{i-1}^3 - K_{i-2}^3}$$

Then the program just iterates. If it doesn't converge, it simply tries a smaller value of  $\lambda$ .

#### fixed\_lambda\_M\_iteration

Arguably the most important function. Takes a value of  $\lambda$  and returns the corresponding K value.

Hard coded into the program are the values of P and M set as

$$\begin{cases} M = 1 \\ P = 0 \end{cases}$$

Sets up spacing for x.  $\tan^2$  spacing is used.

Somewhat concerningly, h' is assumed to already have this spacing, which could potentially cause issues. If you wanted to change the spacing you would have to do it in two different places.

Also a cause for concern, or note is that with this  $\tan^2$  spacing is that the maximum value of  $x_{max}$  is not actually  $x_{max}$  but rather  $x_{max}^2$ . I.e. using  $x_{max}=20$  actually results in the maximum value of x used being 400.

Subroutines then return the kernel matrix & the interpolate matrix. The kernel matrix is in lieu of  $\begin{pmatrix} p \\ 0 \end{pmatrix} = \int \underline{\underline{K}} \begin{pmatrix} g' \\ h' \end{pmatrix}$ . I am not sure of what the interpolate matrix actually is, or what it's for. Finding this out is a big priority.

The matrix A is set up, which is part kernel, part interpolate matrix, same matrix as described earlier.

The roond statement is testing how conditioned the matrix A is, or how ameniable it is to being numerically inverted.

Then the iteration loop begins. Follows Newton's method for the equation  $f(\mathbf{h}') = A\mathbf{h}'$  and iterates via  $\mathbf{h}'_{new} = \mathbf{h}'_{old} + (A - Df|_{\mathbf{h}'_{old}})^{-1}(f(\mathbf{h}'_{old}) - A\mathbf{h}'_{old})$  Where we already know A. f, Df are provided via hprime\_to\_p and f, Df are called p, dp respectively in the program.