

①

Homotopy continuation

① revision 1D case

$$f(x) = x^d + a_1 x^{d-1} + \dots + a_d$$

aim: compute all zeros of $f(x)$

introduce homotopy: $H(x,t) = t(x^d - 1) + (1-t)f(x)$

thus $H(x,1) = x^d - 1$ and $H(x,0) = f(x)$

inserting $f(x)$ one gets

$$H(x,t) = x^d + (1-t) \sum_{k=1}^d a_k x^{d-k} - t$$

• $H(x,t) = 0$ defines d algebraic curves in $[0,1] \times \mathbb{C}$ with $x = \hat{x}_j(t)$ where

$$\hat{x}_j(1) = e^{2\pi i j/d}, \quad j=0, \dots, d-1 \quad \text{and}$$

$$f(\hat{x}_j(0)) = 0$$

Definition. The homotopy continuation method is a numerical method to compute the zeros $\hat{x}_j(0)$ using the starting "approximation" $\hat{x}_j(1)$ and a numerical algorithm to get close to $\hat{x}_j(0)$.

the partial derivatives of H are

$$\frac{\partial H}{\partial t} = - \sum_{k=1}^d a_k x^{d-k} - 1 \quad \text{and} \quad \frac{\partial H}{\partial x} = dx^{d-1} + (1-t) \sum_{k=1}^{d-1} (d-k) a_k x^{d-k-1}$$

where a path $x(t)$ is differentiable one has

$$\frac{dH(x(t),t)}{dt} = \frac{\partial H}{\partial t}(x(t),t) + \frac{\partial H}{\partial x} \frac{dx(t)}{dt} = 0$$

↳

-(2)-

and thus
$$\frac{dx(t)}{dt} = - \frac{\frac{\partial H}{\partial t}(x(t), t)}{\frac{\partial H}{\partial x}(x(t), t)}$$

i.e.

$$\frac{dx}{dt} = \frac{\sum_{k=1}^d a_k x^{d-k} - 1}{d \cdot x^{d-1} + (t-t) \sum_{k=1}^{d-1} (d-k) a_k x^{d-k-1}} \quad (*)$$

$x(t)$ then admits the integral eqn

$$x(t) = x(1) - \int_t^1 r(x(t), t) dt$$

where $r(x, t)$ is the r.h.s. of $(*)$

• discretize parameter t : $t_e = (M-l)h$

$$t_e = t_{e-1} + h, \quad t_0 = 1, \quad t_M = 0 \quad h = \frac{1}{M}$$

$$l = 0, \dots, M$$

• we get from $(*)$:

$$\tilde{x}^{(j)}(t_e) = \tilde{x}^{(j)}(t_{e-1}) + \int_{t_e}^{t_{e-1}} r(\tilde{x}^{(j)}(t), t) dt$$

• we now approximate $\tilde{x}^{(j)}(t_e) \approx w_e^{(j)}$ using a predictor-corrector approach:

(1) predictor gives $\hat{w}_e^{(j)}$ using numerical integrator

(2) use Newton's method with start $\hat{w}_e^{(j)}$ to get $w_e^{(j)}$ (corrector)

- approximate $\tilde{x}^{(j)}(t_n) \approx \hat{w}_e^{(j)}$ by one-step method

$$\hat{w}_e^{(j)} = \text{approx. of } \tilde{x}^{(j)}(t_e), \quad l = 0, \dots, M$$

$$\hat{w}_e^{(j)} = w_{e-1}^j + h \Phi(w_{e-1}^j, t_{e-1}; h)$$

- Euler's method $\Phi(w_{e-1}^j, t_{e-1}; h) = r(w_{e-1}^j, t_{e-1})$

- aim of this method:
find $\hat{w}_e^{(j)}$ such that Newton's method is convergent when starting with $\hat{w}_e^{(j)}$

- theory & numerics of ODE solving relies on Lipschitz continuity (in x) of rhs of ODE

$$\frac{dx}{dt} = r(x, t) \quad \text{see } \textcircled{*}$$

here $r(x, t) = - \frac{\partial H / \partial t}{\partial H / \partial x}$ is rational fct

- problems occur when

$$\frac{\partial H}{\partial x} = d \cdot x^{d-1} + (t-t) \sum_{k=1}^{d-1} (d-k) a_k x^{d-k-1} = 0$$

→ later

Solving $H(x, t) = 0$ with Newton's method

$$v_{k+1} = v_k - \frac{H(v_k, t)}{\frac{\partial H}{\partial x}(v_k, t)}$$

◦ Lipschitz constant: $L = \frac{H(v_k, t)}{\frac{\partial H}{\partial x}(v_k, t) \cdot \frac{\partial^2 H}{\partial x^2}(v_k, t)}$

L is small if v_k close to zero of $H(x, t)$
 and $\frac{\partial H}{\partial x} \cdot \frac{\partial^2 H}{\partial x^2}$ bounded away from zero
 e.g. if zero is simple

◦ this defines the corrector, typically a handful of iterations suffice to get accurate results

◦ start at $\hat{w}_e^{(j)}$ and the final v_k we use as

$$w_e^{(j)} = v_k$$

◦ floating point errors do slow down convergence and lead to an error at the end for ill-conditioned problems.

◦ ill-conditioning in f occurs for t close to 0
 → maybe stop at an earlier t to get some idea of the solution

(t acts a bit like a regularisation parameter)