

Fixed-point problem

Given $G: \mathbb{R}^n \rightarrow \mathbb{R}^n$ find x_* such that $x_* = G(x_*)$

Note: $x_* = G(x_*) \iff F(x_*) = 0$, $F(x) = x - G(x)$

Fixed point iteration*:

```

Given x
while termination do
    x ← G(x)

```

* a.k.a. functional iteration, sometimes Picard iteration
successive substitution

Questions: When does x_* exist?

When is x_* unique?

When does fixed point iteration converge?

(globally? locally?)

How fast does it converge?

- Suppose $\|\cdot\|$ is a norm on $\mathbb{R}^n$, inducing $\|\cdot\|$ on $\mathbb{R}^{n \times n}$
- Define G is a contraction mapping on $D \subseteq \mathbb{R}^n$ if $\exists \lambda$, $0 \leq \lambda < 1$ such that $\|G(x) - G(y)\| \leq \lambda \|x - y\|$ for $x, y \in D$
- see: G a contraction on $D \iff G$ Lipschitz on D with constant $\lambda < 1$
- Also, G is a contraction on D if G is continuously differentiable on D with $\|G'(x)\| \leq \lambda < 1$ for $x \in D$

Why?

$$\begin{aligned}
 G(y) - G(x) &= \int_0^1 \frac{d}{dt} G(x_t) dt, \quad x_t = x + t(y-x) \\
 &= \int_0^1 G'(x_t)(y-x) dt
 \end{aligned}$$

$$\Rightarrow \|G(y) - G(x)\| \leq \left\{ \int_0^1 \|G'(x_t)\| dt \right\} \|y - x\| \leq \lambda \|y - x\|$$

Theorem 1: Suppose G is a contraction mapping on closed $D \in \mathbb{R}^n$. Then $\exists$ unique fixed-point $x_* \in D$ and for any $x_0 \in D$, the fixed point iterates $\{x_k\}$ converge to x_* with

$$\|x_{k+1} - x_*\| \leq \lambda \|x_k - x_*\|$$

proof:

$$\begin{aligned} \text{see: } \|x_{k+1} - x_k\| &= \|G(x_k) - G(x_{k-1})\| \leq \lambda \|x_k - x_{k-1}\| \\ &\leq \lambda^2 \|x_{k-1} - x_{k-2}\| \\ &\leq \dots \end{aligned}$$

$$\Rightarrow \|x_{k+1} - x_k\| \leq \lambda^{k-1} \|x_1 - x_0\|$$

$$\begin{aligned} \Rightarrow \|x_{k+\epsilon} - x_k\| &\leq \|x_{k+\epsilon} - x_{k+\epsilon-1}\| + \|x_{k+\epsilon-1} - x_{k+\epsilon-2}\| + \dots + \|x_{k+1} - x_k\| \\ &\leq \left\{ \lambda^{\epsilon-1} + \lambda^{\epsilon-2} + \dots + \lambda + 1 \right\} \|x_{k+1} - x_k\| \\ &\leq \frac{1}{1-\lambda} \|x_{k+1} - x_k\| \\ &\leq \frac{\lambda^{k-1}}{1-\lambda} \|x_1 - x_0\| \end{aligned}$$

- So by taking k sufficiently large, can make $\|x_{k+\epsilon} - x_k\|$ as small as desired for every $\epsilon > 0$

i.e. $\{x_k\}$ is Cauchy

- Since D is closed (and $\mathbb{R}^n$ is complete^{*}) $\exists x_* \in D$ such that $x_k \rightarrow x_*$

* complete: every Cauchy sequence is convergent

$$\text{see: } G(x_*) = \lim_{k \rightarrow \infty} G(x_k) = \lim_{k \rightarrow \infty} x_{k+1} = x_*$$

$$\text{Also, } \|x_{k+1} - x_*\| = \|G(x_k) - G(x_*)\| \leq \lambda \|x_k - x_*\|$$

Finally, if $\tilde{x}_* \in D$ is a fixed point then

$$\|\tilde{x}_* - x_*\| = \|G(\tilde{x}_*) - G(x_*)\| \leq \lambda \|\tilde{x}_* - x_*\|$$

$$\Rightarrow \tilde{x}_* - x_* = 0 \quad \text{since } \lambda < 1$$

Theorem 2: Suppose G is continuously differentiable near x_* with $\|G'(x_*)\| < 1$. Then for any η with $\|G'(x_*)\| < \eta < 1$, $\exists \delta > 0$ such that if $\|x_0 - x_*\| \leq \delta$, then fixed point iterates $\{x_k\}$ converge to x_* with

$$\|x_{k+1} - x_*\| \leq \eta \|x_k - x_*\|$$

proof:

$$\underbrace{\left[\begin{array}{c} | \\ x_* \end{array} \right]}_{\|G'\| \leq \eta}$$

choose δ such that $\|G'(x)\| \leq \eta$ whenever $\|x - x_*\| \leq \delta$. Set $H_\delta = \{x : \|x - x_*\| \leq \delta\}$

see: H_δ is closed and

$$\begin{aligned} \text{i) } x, y \in H_\delta &\Rightarrow \|G(x) - G(y)\| = \left\| \int_0^1 G'(x_t) dt \right\| \|x - y\| \\ &\leq \eta \|x - y\| \\ &\neq \int_0^1 \|G'(x_t)\| dt \|x - y\| \end{aligned}$$

ii) in particular $x \in H_\delta$

$$\begin{aligned} \Rightarrow \|G(x) - x_*\| &= \|G(x) - G(x_*)\| \\ &\leq \eta \|x - x_*\| \leq \delta \end{aligned}$$

Dissatisfying: Convergence on $\mathbb{R}^n$ is norm independent.
 i.e., since $\mathbb{R}^n$ is finite-dimensional, all norms are equivalent**, so convergence in any norm $\Rightarrow$ convergence in every other norm

** if $\|\cdot\|, \|\cdot\|$ are norms on $\mathbb{R}^n$, then

$$\alpha_1 \|v\| \leq \|v\| \leq \alpha_2 \|v\|$$

α_1, α_2 fixed for any vector v

However, our sufficient condition $\|G'(x_*)\|$ is norm-dependant

For $M \in \mathbb{R}^{n \times n}$, denote

$$\sigma(M) = \{ \lambda : Mx = \lambda x \text{ for some } x \neq 0 \} \quad (\text{spectrum of } M)$$

$$\rho(M) = \max_{\lambda \in \sigma(M)} |\lambda| \quad (\text{spectral radius of } M)$$

Note: For any induced $\|\cdot\|$ on $\mathbb{R}^{n \times n}$ have $\rho(M) \leq \|M\|$

Why? For $\lambda \in \sigma(M)$, $Mx = \lambda x$ for some $x \neq 0$

$$\Rightarrow \|Mx\| = |\lambda| \|x\|$$

$$\Rightarrow |\lambda| = \frac{\|Mx\|}{\|x\|} \leq \|M\|$$

sometimes $\|M\| = \rho(M)$ e.g. if $M = M^T$ and $\|\cdot\| = \|\cdot\|_2$

$$M = U \Lambda U^T \quad \text{where} \quad \Lambda = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} \quad \text{and} \quad U^T U = I$$

and can show

$$\|M\| = \|\Lambda\| = \max |\lambda_i| = \rho(M)$$

Linear algebra fact: Given any $M \in \mathbb{R}^{n \times n}$ and $\epsilon > 0$

$\exists$ induced $\|\cdot\|$ such that

$$\|M\| \leq \rho(M) + \epsilon$$

Theorem 3: Suppose x_* such that $x_* = G(x_*)$ and G is continuously differentiable near x_* with $S(G'(x_*)) \equiv S_* < 1$. Then for any η , $S_* < \eta < 1$, there is a $\delta > 0$ and a norm $\|\cdot\|$ such that if $\|x_0 - x_*\| \leq \delta$ then fixed point iterates $\{x_k\}$ converge to x_* with

$$\|x_{k+1} - x_*\| \leq \eta \|x_k - x_*\|, \text{ for each } k$$

Note: if $S(G'(x_*)) \gg 1$ can't expect $x_k \rightarrow x_*$

$$(***) \quad \dots \quad x_{k+1} - x_* = G(x_k) - G(x_*) = G'(x_*) (x_k - x_*) + o(x_k - x_*)$$



$$G(x) = Ax - G$$

$$G'(x) = A$$

$$Ax = \lambda x, \quad \lambda \text{ eigenvalue}$$

$$\begin{aligned} x_{k+1} - x_* &= Ax_{k+1} - G - Ax_* + G \\ &= A(x_{k+1} - x_*) = A^{k+1}(x_1 - x_0) = \lambda^{k+1}(x_1 - x_0) \end{aligned}$$

$$\text{if } \lambda > 1 \Rightarrow x_{k+1} - x_* \not\rightarrow 0 \quad \underline{\underline{\quad}}$$

special case $\lambda = 1$ Have, assuming $x_k \neq x_*$ for every k

$$G'(x_*) \leftarrow \frac{x_{k+1} - x_*}{x_k - x_*} = G'(x_*) + \frac{o(x_k - x_*)}{x_k - x_*}$$

general λ : if $x_k \rightarrow x_*$ and $G'(x_*) = 0$ then

(***) $\Rightarrow$ superlinear convergence

if $\|G'(x) - G'(x_*)\| \leq \gamma \|x - x_*\|$ then $G'(x_*) = 0$

$\Rightarrow$ quadratic convergence

why? $\| \theta(x_k - x_*) \| = \| G(x_k) - G(x_*) - G'(x_*)(x_k - x_*) \|$

$$= \left\| \int_0^1 [G'(x_t) - G'(x_*)] dt (x_k - x_*) \right\|$$

$\uparrow$
 $x_* + t(x_k - x_*)$

$$\leq \frac{\kappa}{2} \|x_k - x_*\|^2$$

for $G'(x_*) = 0 \Rightarrow \|x_{k+1} - x_*\| \leq \frac{\kappa}{2} \|x_k - x_*\|^2$

Special case $\kappa=1$ again:

- if $G'(x_*) = 0$ and G is twice continuously differentiable near x_* , then

$$\begin{aligned} x_{k+1} - x_* &= G(x_k) - G(x_*) = \\ &= G'(x_*)(x_k - x_*) + \frac{G''(x_*)}{2} (x_k - x_*)^2 + \mathcal{O}((x_k - x_*)^3) \end{aligned}$$

$$\Rightarrow \frac{x_{k+1} - x_*}{(x_k - x_*)^2} = \frac{G''(x_*)}{2} + \frac{\mathcal{O}((x_k - x_*)^3)}{(x_k - x_*)^2} \rightarrow \frac{G''(x_*)}{2}$$

Application 1. Suppose have "Newton-like" iteration

$$x_{k+1} = x_k - B(x_k)^{-1} F(x_k)$$

Treat as $x_{k+1} = G(x_k)$

$$G(x) \equiv x - B(x)^{-1} F(x)$$

if B differentiable, then have

$$G'(x) = I - B(x)^{-1} F'(x) + \mathcal{O}(x - x_*)$$

$$\Leftrightarrow G'(x_*) = I - B(x_*)^{-1} F'(x_*)$$

so have local linear convergence if

$$\|I - B(x_*)^{-1} F'(x_*)\| < 1$$

(special case of earlier result)

In particular $B(x) = F'(x)$ then

$$G'(x_*) = I - F'(x_*)^{-1} F(x_*) = 0$$

and set superlinear convergence (if F is twice continuously differentiable, convergence is quadratic)
see: stronger assumption to get quadratic convergence instead of Lipschitz continuity.

Application 2: Consider Backward differentiation formula method for ODE IVP

$$y' = f(t, y)$$

$$y(t_0) = y_0$$

at m^{th} time step of BDF of order 2 have

$y_{m-1}, \dots, y_{m-2}$ (where $y_i \approx y(t_i)$) and want

$y_m \approx y(t_m)$ such that

$$y_m = \tau \Delta t f(t_m, y_m) + a_m$$

$$a_m = \sum_{j=1}^2 \alpha_j y_{m-j}$$

Obvious fixed point iteration:

start with $y_m^{(0)}$ given by explicit "predictor" method

Iterate:

$$y_m^{(k+1)} = G(y_m^{(k)}), \quad G(y) = \tau \Delta t f(t_m, y) + a_m$$

here $\tau \Delta t$ and $\alpha_1, \dots, \alpha_2$ are (fixed, known)

method constraints

$$G'(y) = \tau \Delta t \frac{\partial f}{\partial y}(t_m, y)$$

$$\|G'(y)\| < 1 \quad \text{if} \quad \tau < \frac{1}{\|\Delta t \frac{\partial f}{\partial y}\|}$$

Application 3: Picard iteration for existence and uniqueness of solution of

$$\left. \begin{aligned} y' &= f(t, y) \\ y(0) &= v \end{aligned} \right\} (*)$$

see: solution $y(t)$ satisfies

$$y(t) = v + \int_0^t f(\tau, y(\tau)) d\tau \quad (**)$$

(and conversely)

Suppose want solution $[0, T]$

Picard iteration

given y_0 , continuous on $[0, T]$

iterate termination

$$y_{k+1}(t) = v + \int_0^t f(\tau, y_k(\tau)) d\tau$$

- Can we extend fixed-point theory to say anything about convergence of $\{y_k\}$?
- Assume $f(t, y)$ is continuous in (t, y) and

$$|f(t, y) - f(t, \tilde{y})| \leq |y - \tilde{y}| \quad \left(\begin{array}{l} \text{Lipschitz} \\ \text{continuous} \end{array} \right)$$

where $|\cdot|$ is a norm on $\mathbb{R}^n$

see: Since y_0 is continuous on $[0, T]$ so is each y_k

set: $C[0, T] = \left\{ y : y \text{ defined and continuous on } [0, T] \right\}$

Then $G(y)$ defined by $G(y)(t) = v + \int_0^t f(\tau, y(\tau)) d\tau$

maps $G : C[0, T] \rightarrow C[0, T]$

And y solves $(*)$ and $(**)$ $\Leftrightarrow y = G(y)$

i.e. $y(t) = G(y)(t)$ for $0 \leq t \leq T$

Note: An contraction mapping theorem 1 still holds if D is a Banach space* (even infinite dimensional)

* vector space with norm $\|\cdot\|$ which is complete i.e. if $\{x_k\}$ is Cauchy in $\|\cdot\|$, then $\exists x_*$ such that $x_k \rightarrow x_*$

Need suitable norm on $C[0, T]$

Choose $K > 0$, define

$$\|y\| = \max_{0 \leq t \leq T} e^{-kt} |y(t)| \quad (\text{max norm})$$

can show: $\{y_k\}$ Cauchy in $\|\cdot\| \Rightarrow \{y_k\}$ converges point wise uniformly on $[0, T]$ to $y_* \in C[0, T]$
 $\Rightarrow \|y_k - y_*\| \rightarrow 0$

That is: $C[0, T]$ is complete in $\|\cdot\|$

For y and z in $C[0, T]$ have

$$|G(y)(t) - G(z)(t)| = \left| V + \int_0^t f(\tau, y(\tau)) d\tau - V - \int_0^t f(\tau, z(\tau)) d\tau \right|$$

$$= \left| \int_0^t [f(\tau, y(\tau)) - f(\tau, z(\tau))] d\tau \right|$$

$$\leq \int_0^t |f(\tau, y(\tau)) - f(\tau, z(\tau))| d\tau$$

$$\leq \int_0^t \gamma e^{k\tau} \cdot e^{-k\tau} |y(\tau) - z(\tau)| d\tau$$

$$\leq \left\{ \max_{0 \leq \tau \leq t} e^{-k\tau} |y(\tau) - z(\tau)| \right\} \int_0^t \gamma e^{k\tau} d\tau$$

$$\leq \|y - z\| \frac{\gamma}{k} (e^{kt} - 1)$$

So

$$e^{-kt} |G(y)(t) - G(z)(t)| \leq \frac{\lambda}{k} (1 - e^{-kt}) \|y - z\|$$
$$\leq \frac{\lambda}{k} \|y - z\|$$

$$\|G(y) - G(z)\| = \max_{0 \leq t \leq T} e^{-kt} |G(y)(t) - G(z)(t)|$$
$$\leq \frac{\lambda}{k} \|y - z\|$$

Since $\frac{\lambda}{k} < 1$, $G: C[0, T] \rightarrow C[0, T]$ is a contraction mapping

By Th. 1. G has a unique fixed point $y \in C[0, 1]$
i.e. (*) and (**) have unique solution y