Lec 22: Rootfinding Problem - One Dimension

Fixed Point Iteration

Fixed Point

Definition 1 (Fixed Point)

The real number r is a **fixed point** of the function q if q(r) = r.

• The rootfinding problem f(x) = 0 can always be written as a fixed point problem q(x) = x by, e.g., setting¹

$$g(x) = x - f(x).$$

• The fixed point problem is true at, and only at, a root of f.

If r is a fixed point of g,
then
$$g(r) = r - f(r) = r$$

$$\Rightarrow y - f(r) = y$$

$$\Rightarrow + f(r) = 0$$

$$\Rightarrow$$
 + $f(r) = \tau$

$$\Rightarrow$$
 + $f(r) = r$

¹This is not the only way to transform the rootfinding problem. More on this later.

Fixed Point Iteration

A fixed point problem g(x) = x naturally provides an iteration scheme:

$$\left\{ \begin{array}{ll} x_0 = \text{initial guess} \\ x_{k+1} = g(x_k), \quad k = 0, 1, 2, \dots. \end{array} \right. \tag{fixed point iteration)}$$

• The sequence $\{x_k\}$ may or may not converge as $k \to \infty$.

If g is continuous and $\{x_k\}$ converges to a number r, then r is a fixed point of g. $g(r) = g\left(\lim_{k \to \infty} x_k\right) = \lim_{k \to \infty} g(x_k) = \lim_{k \to \infty} x_{k+1} = r.$ f = f f = f f = f f = f f = f f = f f = f f = f

Fixed Point Iteration Algorithm

```
function x = fpi(q, x0, n)
% FPI x = fpi(q, x0, n)
% Computes approximate solution of q(x)=x
% Input:
   g function handle
% x0 initial guess
  n number of iteration steps
   x = x0;
   for k = 1:n
       x = g(x); iteration step
(replacement updating)
    end
end
```

Examples

initial iterate to

• To find a fixed point of $g(x) = 0.3\cos(2x)$ near 0.5 using fpi:

```
g = 0 (x) 0.3 * \cos(2 * x);
xc = fpi (g, 0.5, 20)
xc = 0.260266319627758

for h_0 # fiteration
```

Not All Fixed Point Problems Are The Same

The <u>rootfinding problem</u> $f(x) = x^3 + x - 1 = 0$ can be transformed to various fixed point problems:

- $g_1(x) = x f(x) = 1 x^3$
- $g_2(x) = \sqrt[3]{1-x}$
- $g_3(x) = \frac{1+2x^3}{1+3x^2}$

•
$$\chi^3 + \chi - 1 = 0$$
 (RF)

 $\chi^{3} = 1 - \chi$ $\chi = \sqrt[3]{1 - \chi} = g_{2}(\chi)$

Note that all $g_j(x) = x$ are equivalent to f(x) = 0. However, not all these find a fixed point of g, that is, a root of f on the computer.

Exercise. Run fpi with g_j and $x_0=0.5$. Which fixed point iterations converge?

$$\chi = g_3(x) = \frac{1 + 2x^3}{1 + 3x^2} \implies \chi(1 + 3x^2) = 1 + 2x^3$$

$$3x^3 + x = 2x^3 + 1$$

$$4^3 + x - 1 = 0 \quad (RF) \quad \sqrt{3}$$

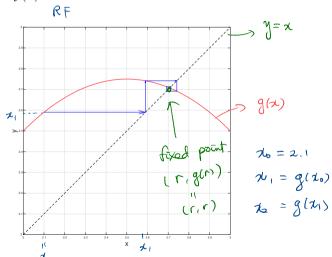
Geometry of Fixed Point Iteration

FP:
$$g(x) = x - f(x)$$

= $-x^2 + 5x - 3.5$

The following script² finds a root of $f(x) = x^2 - 4x + 3.5$ via FPI.

Note the line segments spiral in towards the fixed point.

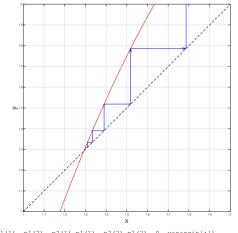


²Modified from FNC.

Geometry of Fixed Point Iteration (cont')

However, with a different starting point, the process does not converge.

```
clf
fplot(g, [1 2], 'r');
hold on
plot([1 2], [1 2], 'k--'),
ylim([1 2])
x = 1.3; y = g(x);
for k = 1:5
    arrow([x y], [y y], 'b');
    x = y; y = g(x);
    arrow([x x], [x y], 'b');
end
```



Series Analysis

Tixed pt (selv)

Let $\epsilon_k=x_k$ the the sequence of errors. (Assume Convergence, i.e. $\epsilon_k \to \infty$)

• The iteration formula $x_{k+1}=g(x_k)$ can be written as

$$\epsilon_{k+1} + r = g(\epsilon_k + r)$$

$$\epsilon_{k+1} + r = g(\epsilon_k + r)$$

$$= g(r) + g'(r)\epsilon_k + \frac{1}{2}g''(r)\epsilon_k^2 + \cdots,$$
 (Taylor series)

implying

$$\epsilon_{k+1} = g'(r)\epsilon_k + O(\epsilon_k^2)$$

assuming sufficient regularity of a.

- Neglecting the second-order term, we have $\epsilon_{k+1} \approx g'(r)\epsilon_k$, which is satisfied if $\epsilon_k \approx C \left[g'(r) \right]^k$ for sufficiently large k.
- Therefore, the iteration converges if |g'(r)| < 1 and diverges if |g'(r)| > 1.

Ourch check: Ext = C[g'(r)] = C g'(r)[g'(r)] = C &

Note: Rate of Convergence

Definition 2 (Linear Convergnece)

Suppose $\lim_{k\to\infty} x_k = r$ and let $\epsilon_k = x_k - r$, the error at step k of an iteration method. If

$$\lim_{k \to \infty} \frac{|\epsilon_{k+1}|}{|\epsilon_k|} = \sigma < 1,$$

the method is said to obey **linear convergence** with rate σ .

Note. In general, say

$$\lim_{k \to \infty} \frac{\left|\epsilon_{k+1}\right|}{\left|\epsilon_{k}\right|^{p}} = \sigma$$

for some $p \ge 1$ and $\sigma > 0$.

• If
$$p=1$$
 and

- $\sigma = 1$, the convergence is *sublinear*;
- $0 < \sigma < 1$, the convergence is linear;
- $\sigma = 0$, the convergence is *superlinear*.
- If p = 2, the convergence is *quadratic*;
- If p = 3, the convergence is *cubic*, ...

Convergence of Fixed Point Iteration

Theorem 3 (Convergence of FPI)

Assume that g is continuously differentiable, that g(r) = r, and that $\sigma = |g'(r)| < 1$. Then the fixed point iterates x_k generated by

$$x_{k+1} = g(x_k), \quad k = 1, 2, \dots,$$

converge linearly with rate σ to the fixed point r for x_0 sufficiently close to r.

In the previous example with $g(x) = x - f(x) = -x^2 + 5x - 3.5$:

- For the first fixed point, near 2.71, we get $g'(r) \approx -0.42$ (convergence);
- For the second fixed point, near 1.29, we get $g'(r) \approx 2.42$ (divergence).

Note. An iterative method is called **locally convergent** to r if the method converges to r for initial guess sufficiently close to r.

Contraction Maps

Lipschitz Condition

A function g is said to satisfy a **Lipschitz condition** with constant L on the interval $S \subset \mathbb{R}$ if

$$|g(s) - g(t)| \le L|s - t|$$
 for all $s, t \in S$.

- A function satisfying the Lipschitz condition is continuous on S.
- If L < 1, g is called a **contraction map**.

When Does FPI Succeed?

Contraction Mapping Theorem

Suppose that g satisfies Lipschitz condition on S with L < 1, i.e., g is a contraction map on S. Then S contains exactly one fixed point r of g. If x_1, x_2, \ldots are generated by the fixed point iteration $x_{k+1} = g(x_k)$, and x_1, x_2, \ldots all lie in S, then

$$|x_k - r| \le L^{k-1} |x_1 - r|, \quad k > 1.$$

Newton's Method

FPI, when convergent, is thearly convergent.

Newton's Method

To find the root of f:

Newton's Method (Algorithm)

• Begin at the point $(x_0, f(x_0))$ on the curve and draw the tangent line at the point using the slope $f'(x_0)$:

$$y = f(x_0) + f'(x_0)(x - x_0).$$

$$y=f(x_0)+f'(x_0)(x-x_0).$$
• Find the x -intercept of the line and call it x_1 :
$$f(x_0)$$

(*****)

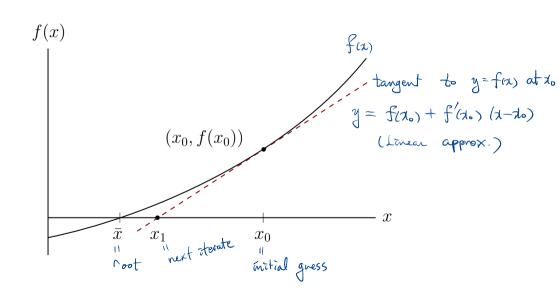
$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} \,.$$

• Continue this procedure to find x_2, x_3, \ldots until the sequence converges to the root.

General iterative formula:

will:
$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} \quad \text{for } k = 0, 1, 2, \dots$$

Newton's Method: Illustration



Series Analysis

Let
$$\epsilon_k = x_k - r, k = 1, 2, \ldots$$
, where r is the limit of the sequence and $f(r) = 0$.

Substituting
$$x_k=r+\epsilon_k$$
 into the iterative formula (*): $-$

$$\epsilon_{k+1} = \epsilon_k - \frac{f(r + \epsilon_k)}{f'(r + \epsilon_k)}.$$

Taylor-expand f about x = r and simplify (assuming $f'(r) \neq 0$):

$$\epsilon_{k+1} = \epsilon_k - \frac{f(r) + \epsilon_k f'(r) + \frac{1}{2} \epsilon_k^2 f''(r) + O(\epsilon_k^3)}{f'(r) + \epsilon_k f''(r) + O(\epsilon_k^2)}$$

$$= \epsilon_k - \epsilon_k \left[1 + \frac{1}{2} \frac{f''(r)}{f'(r)} \epsilon_k + O(\epsilon_k^2) \right] \left[1 + \frac{f''(r)}{f'(r)} \epsilon_k + O(\epsilon_k^2) \right]^{-1}$$

$$= \frac{1}{2} \frac{f''(r)}{f'(r)} \epsilon_k^2 + O(\epsilon_k^3).$$

$$= \frac{1}{2} \frac{f''(r)}{f'(r)} \epsilon_k^3 + O(\epsilon_k^3).$$

Exercise. Check algebra.

G €K+1 2 C €K