

Fixed Point Iteration

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Fixed Point

Definition 1 (Fixed Point)

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

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

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

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

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:

$$\begin{cases} x_0 = \text{initial guess} \\ x_{k+1} = g(x_k), \quad k = 0, 1, 2, \dots \end{cases} \quad (\text{fixed point iteration})$$

- The sequence $\{x_k\}$ may or may not converge as $k \rightarrow \infty$.
- If g is continuous and $\{x_k\}$ converges to a number r , then r is a fixed point of g .

$$g(r) = g\left(\lim_{k \rightarrow \infty} x_k\right) = \lim_{k \rightarrow \infty} g(x_k) = \lim_{k \rightarrow \infty} x_{k+1} = r.$$

Fixed Point Iteration Algorithm

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

Examples

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

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

```
xc = 0.260266319627758
```

Not All Fixed Point Problems Are The Same

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

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

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.

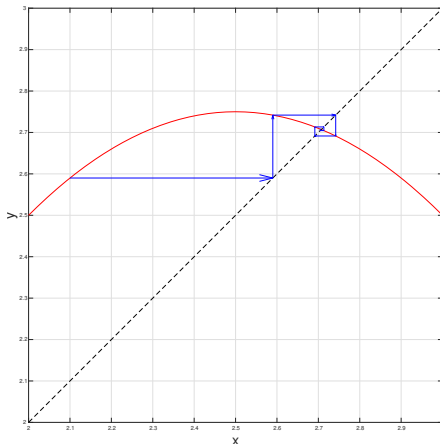
Exercise. Run `fpi` with g_j and $x_0 = 0.5$. Which fixed point iterations converge?

Geometry of Fixed Point Iteration

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

```
f = @(x) x.^2 - 4*x + 3.5;  
g = @(x) x - f(x);  
fplot(g, [2 3], 'r');  
hold on  
plot([2 3], [2 3], 'k--')  
x = 2.1;  
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
```

Note the line segments spiral in towards the fixed point.

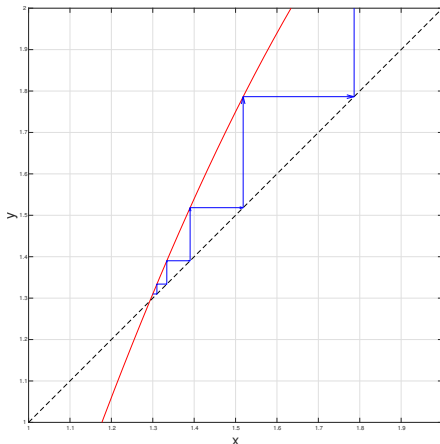


²Modified from FNC.

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
```



Custom function: `arrow = @(p1, p2, varargin) quiver(p1(1), p1(2), p2(1)-p1(1), p2(2)-p1(2), 0, varargin{:})`

Series Analysis

Let $\epsilon_k = x_k - r$ be the sequence of errors.

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

$$\begin{aligned}\epsilon_{k+1} + r &= g(\epsilon_k + r) \\ &= g(r) + g'(r)\epsilon_k + \frac{1}{2}g''(r)\epsilon_k^2 + \cdots, \quad (\text{Taylor series})\end{aligned}$$

implying

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

assuming sufficient regularity of g .

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

Note: Rate of Convergence

Definition 2 (Linear Convergnece)

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

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

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

Note. In general, say

$$\lim_{k \rightarrow \infty} \frac{|\epsilon_{k+1}|}{|\epsilon_k|^p} = \sigma$$

for some $p \geq 1$ 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

$$|g(s) - g(t)| \leq L |s - t| \quad \text{for all } s, t \in S.$$

- 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, \dots are generated by the fixed point iteration $x_{k+1} = g(x_k)$, and x_1, x_2, \dots all lie in S , then

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