

Fixed point iterations for real functions

Motivation example: the logistic map – a very simple model of evolution of a population

$x_{n+1} = r x_n (1 - x_n)$, where x_n is a number in $(0, 1)$

– x_n represents the ratio of existing population to the maximum possible one, at n -th year

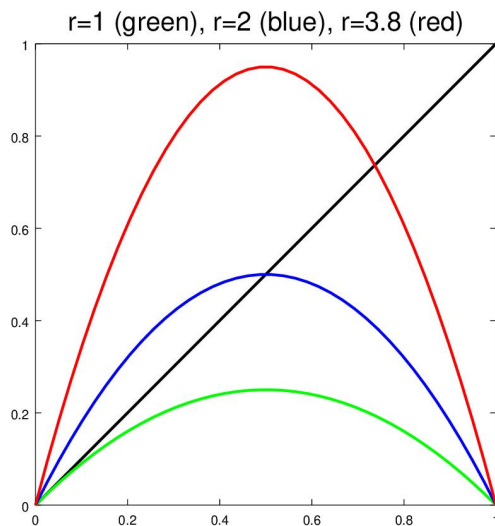
– r is a parameter of the model ($0 < r < 4$, so that if $x_n \in (0, 1)$, then $x_{n+1} \in (0, 1)$)

Question 1: Does there exist a *fixed point* x^* that solves the *logistic equation*

$$x^* = r x^* (1 - x^*) ?$$

– x^* represents some steady state, or equilibrium

The graphical solution:



Question 2: Does the logistic map (the "evolution") converge to a fixed point?

This is a special example of a more general formulation that will be discussed in this text.

Fixed point iterations

The goal: find the solution of a fixed point equation

$$x = g(x), \quad \text{where } x \in R, \quad g : R \rightarrow R.$$

Fixed point iterations (fpi): choose x_0 , then for $k = 0, 1, 2, \dots$ compute

$$x_{k+1} = g(x_k)$$

Example 1: $g(x) = 0.2x^2 + 1$, use $x_0 = 0.2$ and precision up to 4 decimal places

$$x_1 = g(x_0) = 0.2x_0^2 + 1 = 0.2 \cdot 0.2^2 + 1 = 0.0080 + 1 = 1.0080$$

$$x_2 = g(x_1) = 0.2x_1^2 + 1 = 0.2 \cdot 1.0080^2 + 1 = 1.2032$$

$$x_3 = g(x_2) = 0.2x_2^2 + 1 = 0.2 \cdot 1.2032^2 + 1 = 1.2895$$

$$x_4 = 0.2x_3^2 + 1 = 0.2 \cdot 1.2895^2 + 1 = 1.3326$$

...

$$x_{13} = 0.2x_{12}^2 + 1 = 1.3818$$

$$x_{14} = 0.2x_{13}^2 + 1 = 1.3819$$

$$x_{15} = 0.2x_{14}^2 + 1 = \underbrace{0.2 \cdot 1.3819^2 + 1}_{g(1.3819)} = 1.3819$$

$g(1.3819) = 1.3819 \Rightarrow x = g(x)$ holds with accuracy up to 4 decimal places

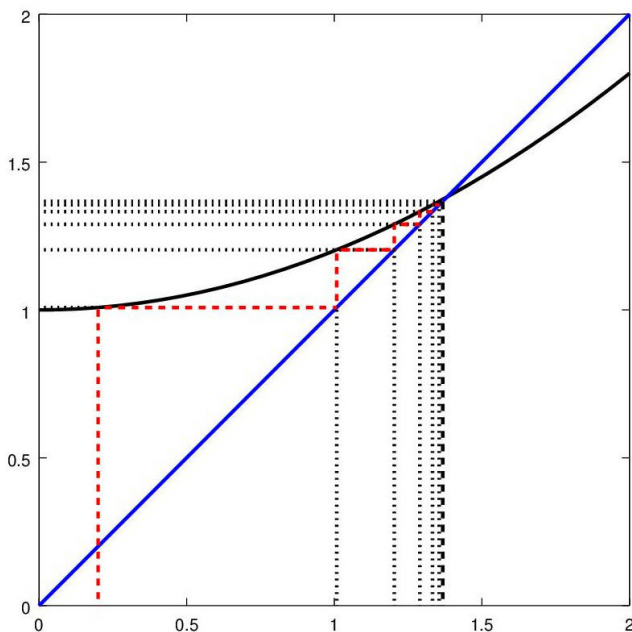


Figure 1: Graphical illustration of iterations $x_{k+1} = g(x_k)$: black: $y = g(x) = 0.2x^2 + 1$, blue: $y = x$.

Graphical solution – the algorithm:

- choose x_0 and set $P_0 = [x_0, x_0]$
- for $k = 0, 1, 2, \dots$, repeat
 1. move vertically to the curve $y = g(x)$: set $Q_k = [x_k, g(x_k)]$ and $x_{k+1} = g(x_k)$
 2. move horizontally to the line $y = x$: set $P_{k+1} = [x_{k+1}, x_{k+1}]$
 until $|x_{k+1} - x_k| < \text{some given precision}$

The resulting broken line $P_0, Q_0, P_1, Q_1, \dots, P_k, Q_k, \dots$ (red, dashed) converges to the graphical solution.

Question: under which assumptions fpi converge (i.e. the loop in the algorithm above is finite)?

Example 2: $g(x) = 0.2x^2 + 1$ as in ex. 1; use $x_0 = 3.4$ (below left) and then $x_0 = 3.8$ (below right)

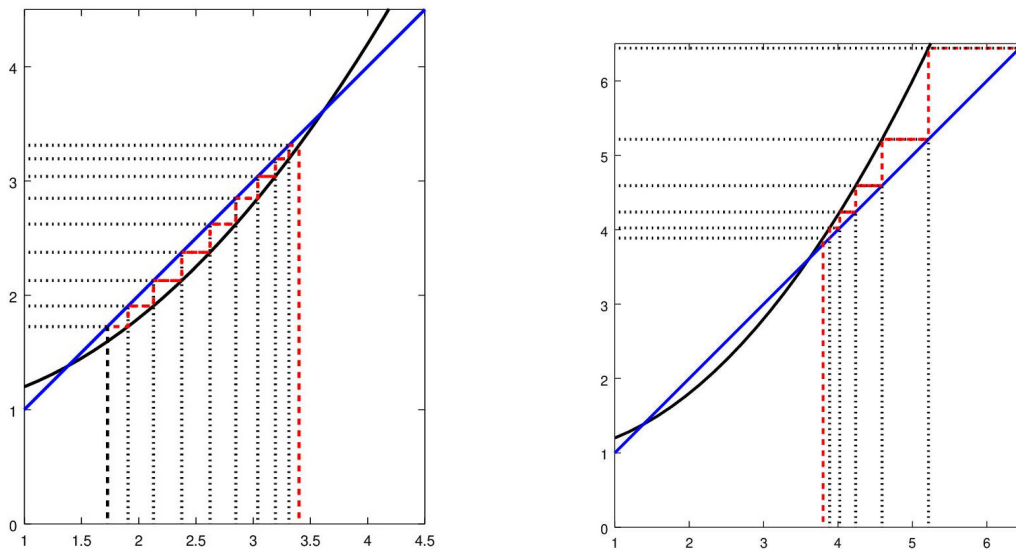


Figure 2: Ex. 2 – convergent staircase, to the left solution (left), divergent staircase (right).

Example 3: $g(x) = \cos x$, use $x_0 = 0.4$ (below left)

Example 4: $g(x) = -1.4(x - 2) + 0.8 \cos x + 20$, use $x_0 = 8.9$ (below right)

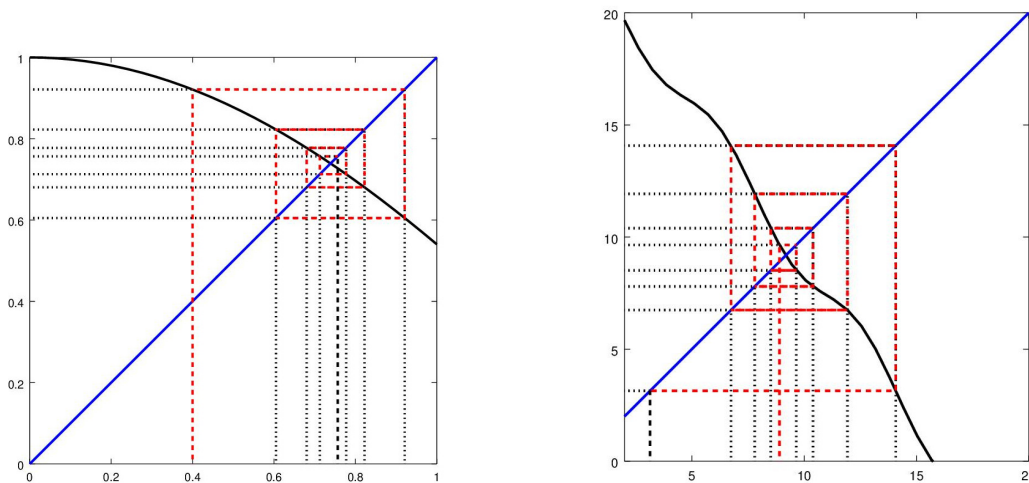


Figure 3: Left: ex. 3 – spiral inwards (convergent). Right: ex. 4 – spiral outwards (divergent).

Observations:

- $g(x)$ is increasing \Rightarrow staircase (ex. 1, 2, Figures 1, 2),
 $g(x)$ is decreasing \Rightarrow spiral (ex. 3, 4, Figure 3).
- the slope of $g(x)$ is small \Rightarrow fpi converge (ex. 1, 2, 3, left figures),
the slope of $g(x)$ is large, steep \Rightarrow fpi diverge (ex. 2, 4, right figures)
the treshold $\approx 45^\circ$

Theorem 1 – contraction mapping theorem in R:

Let $D \subset R$, D be closed, let $g : D \rightarrow R$ be a real function. Assume

1. if $x \in D$, then $g(x) \in D$
2. the function g is a *contraction* on D : there exists a number $q < 1$ such that

$$|g(x) - g(y)| \leq q \cdot |x - y| \quad \forall x, y \in D \tag{1}$$

Then

- there exists unique $x^* \in D$ such that $x^* = g(x^*)$ and the fixed point iterations converge to x^* for any choice of $x_0 \in D$,
- x_k satisfies the a-priori error estimate $|x_k - x^*| \leq \frac{q^k}{1-q} |x_1 - x_0|$
and the a-posteriori error estimate $|x_k - x^*| \leq \frac{q}{1-q} |x_k - x_{k-1}|$.

Proof: see [1]

Theorem 2 – the sufficient condition for the contraction property in R:

Let $g(x)$ be a differentiable function on interval $I \subset R$, such that $|g'(x)| \leq q < 1 \quad \forall x \in I$. Then g is a contraction in I and satisfies (1), i.e. $|g(x) - g(y)| \leq q \cdot |x - y| \quad \forall x, y \in I$.

Proof: from Taylor’s expansion for function $g(x)$, we have $\forall x, y \in I$:

$$\begin{aligned} g(y) &= g(x) + g'(\xi) \cdot (x - y) , \quad \text{where } \xi \in (x, y) \\ g(y) - g(x) &= g'(\xi) \cdot (x - y) \\ |g(x) - g(y)| &= |g'(\xi)| \cdot |x - y| \leq q \cdot |x - y| \end{aligned}$$

Fixed point iterations for general equation $f(x) = 0$

Transform the equation to the previous form:

$$\begin{aligned} 0 &= f(x) \quad / \cdot \alpha \\ 0 &= \alpha f(x) \\ x &= \underbrace{x + \alpha f(x)}_{g(x)} \end{aligned}$$

The coefficient α has to be chosen so that $g(x) = x + \alpha f(x)$ is a contraction.

Example 5: solve numerically $2 \operatorname{arctg} x + \log x = 0$.

The function $f(x) = 2 \operatorname{arctg} x + \log x$ is defined for $x \in (0, \infty)$, values of $f(x)$ are growing from $-\infty$ to ∞ , so the equation has unique solution x^* . In this case, the solution can be easily located to, say, $x^* \in (0.3, 1)$, as $f(0.3) < 0$ and $f(1) > 0$.

The function $f(x)$ is differentiable – we will try to find α so that the assumptions of Theorem 2 hold:

$$|g'(x)| = |1 + \alpha f'(x)| \leq q < 1 \text{ for } x \in I = (0.3, 1) \text{ and } f(I) \subset I.$$

Let us use some guesswork: $f'(x) > 0$, so α has to be some negative number, small enough so that $1 + \alpha f'(x) > -1$. Choose $x_0 = 1$ and examine some values of α . We can check that values $\alpha = -0.1$, or $\alpha = -0.2$, or $\alpha = -0.3$ work perfectly ($x^* \approx 0.438$ after few iterations), value $\alpha = -0.4$ is not so good and for $\alpha < -0.5$ the fpi method does not converge.

References

- [1] Tobias von Petersdorff: Contraction mapping theorem
<http://terpconnect.umd.edu/~petersd/666/fixedpoint.pdf>