

# 5. Fixed Point Method

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Lecture Note for MAT325 Numerical Analysis

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## 1 Introduction

The number  $p$  is a fixed point for a given function  $g$  if  $g(p) = p$ . In other words, if function  $g(x)$  has a fixed point  $p$ , then  $p$  is a root of equation  $g(x) - x = 0$ . Conversely, we could convert root finding problem  $f(x) = 0$  to fixed-point problem  $g(x) = af(x) - x = x$  for any real number  $a \neq 0$ . This section discusses how to approximate the root  $p$  using the **Fixed-point Method**.

## 2 Some Theories of the Fixed Point Method

**Theorem 1** (Existence of fixed point)

Suppose that  $f(x)$  is a continuous function that maps its domain,  $D$ , onto a subset of itself,  $S = f(D)$ , i.e.,  $f(x) \in C[a, b]$ , such that

$$f(x) : [a, b] \in S \subset [a, b]$$

Then  $f(x)$  has a fixed point in  $[a, b]$ .

**Proof:** If  $f(a) = a$  or  $f(b) = b$ , we are done.

Assume  $f(a) \neq a$  and  $f(b) \neq b$ . Since  $f([a, b]) \subset [a, b]$ , we have  $f(a) > a$  and  $f(b) < b$ . Let  $h(x) = f(x) - x$ . Since  $f(x) \in C[a, b]$ , so is  $h(x) \in C[a, b]$ . Observe that  $h(a) = f(a) - a > 0$ ,  $h(b) = f(b) - b < 0$ . By intermediate value theorem, there exists at least one value  $r$  in  $[a, b]$  such that  $h(r) = 0$ . That implies that  $f(r) = r$ . This completes the proof.

**Corollary:** Every continuous bounded function on the real numbers has a fixed point.

**Theorem 2** (Uniqueness of fixed point, also called Contraction Mapping Theorem)

Suppose that  $f(x)$  is a continuous function that maps its domain,  $D$ , onto a subset of itself,  $S = f(D)$ , i.e.,  $f(x) \in C[a, b]$ , such that

$$f(x) : [a, b] \in S \subset [a, b]$$

Suppose further that there exists some positive constant  $0 < K < 1$  such that  $|f'(x)| < K$  for all  $x$  in  $[a, b]$ . Then  $f(x)$  has a unique fixed point  $p$  in  $[a, b]$ .

**Proof:** Assume there at least two fixed points, say  $p_1$  and  $p_2$  in  $[a, b]$ , such that  $f(p_1) = p_1$  and  $f(p_2) = p_2$  with  $p_1 < p_2$ . That is,  $f(p_2) - f(p_1) = p_2 - p_1$ .

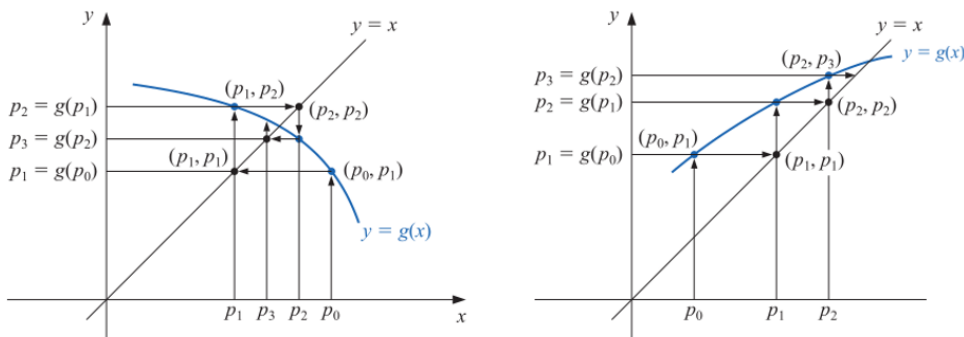
On the other hand, the mean value theorem indicated that  $f(p_2) - f(p_1) = f'(c)(p_2 - p_1)$  for some  $c$  in  $[p_1, p_2]$ . Since, we have  $f(p_2) - f(p_1) < p_2 - p_1$ . This **contradicts** with  $f(p_2) - f(p_1) = p_2 - p_1$ . Therefore, we have proved the uniqueness of the fixed point. The proof is completed.

### 3 Fixed Point Iteration

With the theory developed previously, we focus on the following recursive equation for finding the fixed point numerically.

Choose an initial approximation  $p_0$ , generate sequence  $\{p_n\}_{n=0}^{\infty}$  by  $p_n = g(p_{n-1})$ . If the sequence converges to  $p$  and  $g(x)$  is continuous at  $p$ , then

$$p = \lim_{n \rightarrow \infty} p_n = \lim_{n \rightarrow \infty} g(p_{n-1}) = g(p)$$



The following animated graph demonstrates the process of searching the fixed point.

[https://github.com/pengdsci/MAT325/raw/main/w03/img/Fixed\\_point\\_anime.gif](https://github.com/pengdsci/MAT325/raw/main/w03/img/Fixed_point_anime.gif)

#### Fixed-point Iteration Algorithm

```

INPUT    f(x),
         initial p0,
         TOL
         N                               (max iterations)
STEP 1.  ERR = infinity                  (initial error - must be a big number)
         x                               (initial value)
         n = 0                           (initial iterator)
STEP 2.  WHILE ERR > TOL DO:
STEP 3.  n = n + 1
STEP 4.  new.x = f(x)
         ERR = new.x - x
         IF ERR < TOL DO:
           OUTPUT (results)

```

```

        STOP
    ENDIF
IF ERR > TOL DO:
    OUTPUT (intermediate results)
    x = new.x      (update x value)
ENDIF
IF n == N DO:
    OUTPUT ( )
    STOP
ENDIF

```

## 4 Error Analysis

By the Lipschitz condition in the definition of **Contraction Mapping Theorem**, the approximated error is easily obtained as  $|p_{n+1} - p_n| \leq K^n |p_1 - p_0|$ . Next, we present a result that gives the bound the true error.

**Theorem 3** (Error Estimation)

If fixed point iteration is terminated after  $n > 1$  steps then the error is limited by

$$|p_n - p| \leq \frac{K^n |p_1 - p_0|}{1 - K}$$

where  $0 < K < 1$  is the Lipschitz constant.

**Proof** We use mathematical induction to prove this theorem. For  $n = 1$ , we need to show that

$$|p_1 - p| \leq \frac{K}{K - 1} |p_1 - p_0|.$$

To this end, we use the Mean Value Theorem, there exists a constant  $c \in [\min\{p, p_0\}, \max\{p, p_0\}]$  such that

$$|g'(c)| = \left| \frac{f(p_0) - f(p)}{p_0 - p} \right| = \left| \frac{p_1 - p}{p_0 - p} \right| \leq K.$$

That is,  $|p_1 - p| \leq K |p_0 - p|$ . Using the triangular inequality, we have

$$|p_1 - p| \leq K |p_0 - p| = K |p_0 - p_1 + p_1 - p| \leq K |p_1 - p_0| + K |p_1 - p|$$

Therefore

$$|p_1 - p| \leq \frac{K}{K - 1} |p_1 - p_0|.$$

We assume that original inequality is true when  $n = m$ , i.e.,

$$|p_m - p| \leq \frac{K^m |p_1 - p_0|}{1 - K}.$$

We want to show that

$$|p_{m+1} - p| \leq \frac{K^{m+1} |p_1 - p_0|}{1 - K}.$$

We still use the MVT on interval  $[\min\{p, p_m\}, \max\{p, p_m\}]$  to get

$$|g'(c)| = \left| \frac{f(p_m) - f(p)}{p_m - p} \right| = \left| \frac{p_{m+1} - p}{p_m - p} \right| \leq K.$$

where  $c \in [\min\{p, p_m\}, \max\{p, p_m\}]$ . Therefore,

$$|p_{n+1} - p| \leq K|p_n - p|$$

and

$$|p_{m+1} - p| \leq K|p_m - p_0| = \frac{K^{m+1}}{1-L}|p_1 - p_0|.$$

This completes the proof.

**Corollary** (Order of Convergence)

If the fixed-point method converges, it has a linear convergence order.

The proof has already been given in the proof of the above theorem.

## 5 Numerical Example

We know that there is a solution for the equation  $x^3 - 7x + 2 = 0$  in  $[0, 1]$ . We rewrite the equation in the form  $x = (x^3 + 2)/7$  and denote the process  $x_{n+1} = (x_n^3 + 2)/7$ . We see from the following figures that if  $0 \leq x_0 \leq 1$  then  $(x_n)$  converges to a root of the above equation. We also note that if we start with (for example)  $x_0 = 2.5$  then the recursive process does not converge.

```
## input values
TOL = 10^(-6)
M = 200
####
gfun = function(x) (x^3 + 2)/7
####
a = 0
b = 2
x = 2      # initial value of x
n = 0      # initializing iterator
ERR = Inf  # initial error
##
while (ERR > TOL){
  n = n + 1
  new.x = gfun(x)
  ERR = abs(new.x - x)
  if(ERR < TOL){
    cat("\n\nThe algorithm converges!")
    cat("\nThe approximate root is:", new.x, ".")
    cat("\nThe absolute error is:", ERR, ".")
    cat("\nThe number of iterations is:", n, ".")
    break
  } else{
    if(ERR > 10^7){
      cat("\n\nThe algorithm diverges!")
      break
    } else{
```

```

        cat("\nIteration:",n,". Estimated root:", new.x, ". Absolute error:", ERR, ".")
        x = new.x          # update x value!!!
    }
}
if(n == M){
    cat("\n\nThe maximum number of iterations is achieved!")
    break
}
}

```

```

##
## Iteration: 1 . Estimated root: 1.428571 . Absolute error: 0.5714286 .
## Iteration: 2 . Estimated root: 0.7022074 . Absolute error: 0.726364 .
## Iteration: 3 . Estimated root: 0.3351793 . Absolute error: 0.3670281 .
## Iteration: 4 . Estimated root: 0.2910937 . Absolute error: 0.04408562 .
## Iteration: 5 . Estimated root: 0.289238 . Absolute error: 0.001855685 .
## Iteration: 6 . Estimated root: 0.289171 . Absolute error: 6.696095e-05 .
## Iteration: 7 . Estimated root: 0.2891686 . Absolute error: 2.400242e-06 .
##
## The algorithm converges!
## The approximate root is: 0.2891685 .
## The absolute error is: 8.601697e-08 .
## The number of iterations is: 8 .

```

## 6 Chapter 2 Homework Part II

Section 2.2:

Problem 1. Page 18.

Problem 7. Page 18 (first part): Use Theorem 2.3 to show that  $g(x) = \pi + 0.5 \sin(x/2)$  has a unique fixed point on  $[0, 2\pi]$ .

Problem 12(a). Page 19.