

Class 12: September 23, 2024

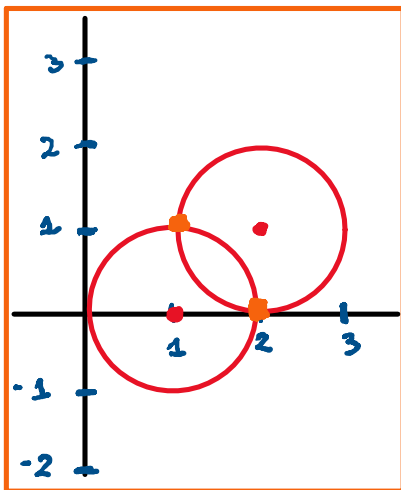
Recall: Last time, we wrapped up our first unit on rootfinding methods aimed at solving one non-linear equation in one real variable. At the very beginning of the unit, we argued that any non-linear equation like $x + \cos(x) = 3$ can be easily turned into a problem of the form: *find x such that $f(x) = 0$* . We then spent some time developing and analyzing the following algorithms:

1. **Bisection method:** Assuming f is $C[a,b]$ and changes sign, it is guaranteed to converge linearly with rate ~ 0.5 .
2. **Fixed Point Iteration:** If we can find $g(x)$ such that $g(x) = x$ is equivalent to $f(x) = 0$, we can apply FPI to find the roots of $f(x)$. Assuming g is continuous, $g(x)$ in $[a,b]$ and is "contractive" around the root(s), it converges at least linearly (can be higher order, e.g. quadratic!). You can usually only get guarantees for x_0 close enough to r .
3. **Newton-Raphson method:** Uses access to $f'(x)$ to define a step as the root of the tangent line at the previous guess. It gives us a very special FPI method that, for x_0 close enough to r , converges quadratically. Very powerful, but unreliable unless we can pair it with something else.
4. **Secant method:** Imitates the best features of Newton while avoiding computing $f'(x)$. It uses two initial guesses x_0 and x_1 to define the secant line, the step is then the root of the secant line using the latest 2 iterations. For x_0 close enough to r , converges superlinearly (order $\phi \sim 1.618$).

However, it is rare to just have to solve one equation in one unknown. In applications, more often than not, we will have M non-linear equations with N unknowns, where M and N are potentially large. In this section, we will generalize some of the rootfinding methods discussed above to this case, and discuss how implementation / performance / costs behave in this setting. What is the same as the 1D case? What is different?

SYSTEMS OF NONLINEAR EQS:

EXAMPLE: Find the intersection pts for two circles of radius 1 centered at $(1,0)$ and $(2,1)$.



Two solutions:

$$\vec{r}_1 = (2, 0), \quad \vec{r}_2 = (1, 1).$$

Algebraically:

$$\begin{cases} \text{Eq1. } (x_1 - 1)^2 + x_2^2 = 1 \\ \text{Eq2. } (x_1 - 2)^2 + (x_2 - 1)^2 = 1 \end{cases}$$

SOLUTIONS OF A SYSTEM OF 2 NONLIN EQS IN 2 VARS.

How do we relate this to rootfinding?

DEFINE $F: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$F(\underbrace{[x_1, x_2]}_{\vec{x}}) = \begin{bmatrix} f_1([x_1, x_2]) \\ f_2([x_1, x_2]) \end{bmatrix} = \begin{bmatrix} (x_1 - 1)^2 + x_2^2 - 1 \\ (x_1 - 2)^2 + (x_2 - 1)^2 - 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

s.t. solving our system \leftrightarrow finding \vec{v} such that $\underline{F(\vec{v}) = \vec{0}}$.

In general: $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$, $F(\underline{\vec{x}}) = \underline{\vec{0}}$
 $\in \mathbb{R}^n$ \mathbb{R}^m

m equations $f_i(\vec{x}) = 0$, $i=1, \dots, m$.
in n variables

Assume $m=n$.

FIXED POINT ITERATION

Find G such that $G(\vec{x}) = \vec{x} \leftrightarrow F(\vec{x}) = \vec{0}$.

Candidates: $G(\vec{x}) = \vec{x} + c F(\vec{x})$, $c \neq 0$.

$$G(\vec{v}) = \vec{v}$$

$$\vec{v} = \vec{x} + c F(\vec{v}) \rightarrow c F(\vec{v}) = \vec{0}.$$

$$G(\vec{x}) = \vec{x} + C(\vec{x}) F(\vec{x})$$

$$\begin{bmatrix} g_1(x_1, x_2) \\ g_2(x_1, x_2) \end{bmatrix} + \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + C(\vec{x}) \begin{bmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \end{bmatrix}$$

$$G(\vec{x}) = \vec{x} + \underline{C(\vec{x})} F(\vec{x})$$

2x2 matrix.

$$\vec{v} = G(\vec{v}) = \vec{v} + \underline{C(\vec{v})} F(\vec{v})$$

$$\boxed{\vec{0} = \underline{C(\vec{v})} F(\vec{v})}$$

2x2 2x1

$\underline{C(\vec{x})}$
invertible
around
 \vec{v}

$$\left| \frac{\vec{r}}{r} \right|$$

1D: Condition $|g'(x)| \leq K < 1$ on a neigh of $r \rightarrow$ Linear convergence.

nD: Condition on $\|J_G\| \leq K < 1$?

• for $x, y \in [a, b]$, $|g(x) - g(y)| \leq K |x - y|$, $K < 1$

for \vec{x}, \vec{y} on neigh of r , $\|G(\vec{x}) - G(\vec{y})\| \leq K \|\vec{x} - \vec{y}\|$

$$\|(x_1, x_2)\|_2 = \sqrt{x_1^2 + x_2^2}$$

$$\|(x_1, x_2)\|_\infty = \max(|x_1|, |x_2|)$$

$$\|(x_1, x_2)\|_1 = |x_1| + |x_2|$$

\vec{x}_n close to r ,

$$1D: e_{n+1} = g'(r) e_n + \frac{1}{2} g''(\xi_n) \cdot e_n^2$$

$$nD: \underline{\vec{e}_{n+1} = J_G(\vec{r}) \cdot \vec{e}_n + \text{H.O.T } O(\|\vec{e}_n\|^2)}$$