

APPM 4650 — Fixed point for non-linear systems

Consider the problem of find $\begin{bmatrix} \alpha \\ \beta \end{bmatrix}$ such that

$$\begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \begin{bmatrix} f(\alpha, \beta) \\ g(\alpha, \beta) \end{bmatrix}.$$

As in one dimension, we can create a fixed point iteration

$$\begin{bmatrix} x_{n+1} \\ y_{n+1} \end{bmatrix} = \begin{bmatrix} f(x_n, y_n) \\ g(x_n, y_n) \end{bmatrix} \quad \text{for } n = 0, 1, 2, \dots$$

How do we know when this will converge?

Let's try to mimic what we did for one dimensional problems. Assume f and g are analytic. (This is more than sufficient.)

Let $\begin{bmatrix} x_n \\ y_n \end{bmatrix} = \begin{bmatrix} \alpha + \Delta x_n \\ \beta + \Delta y_n \end{bmatrix}$

Then the fixed point iteration becomes

$$\begin{aligned} \begin{bmatrix} x_{n+1} \\ y_{n+1} \end{bmatrix} &= \begin{bmatrix} f(x_n, y_n) \\ g(x_n, y_n) \end{bmatrix} \\ \begin{bmatrix} \alpha + \Delta x_{n+1} \\ \beta + \Delta y_{n+1} \end{bmatrix} &= \begin{bmatrix} f(\alpha + \Delta x_n, \beta + \Delta y_n) \\ g(\alpha + \Delta x_n, \beta + \Delta y_n) \end{bmatrix} \end{aligned}$$

Now we can Taylor expand the right hand side centering at $\begin{bmatrix} \alpha \\ \beta \end{bmatrix}$. This results in

$$\begin{bmatrix} \alpha \\ \beta \end{bmatrix} + \begin{bmatrix} \Delta x_{n+1} \\ \Delta y_{n+1} \end{bmatrix} = \begin{bmatrix} f(\alpha, \beta) \\ g(\alpha, \beta) \end{bmatrix} + \underbrace{\begin{bmatrix} f_x & f_y \\ g_x & g_y \end{bmatrix}}_{\mathbf{G}}|_{(\alpha, \beta)} \begin{bmatrix} \Delta x_n \\ \Delta y_n \end{bmatrix} + \text{higher order terms}$$

where f_x denotes the partial derivative of f with respect to x , etc.

Since $\begin{bmatrix} \alpha \\ \beta \end{bmatrix}$ is the fixed point $\begin{bmatrix} f(\alpha, \beta) \\ g(\alpha, \beta) \end{bmatrix} = \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$. The matrix \mathbf{G} is called the *Jacobian*.

$$\begin{bmatrix} \alpha \\ \beta \end{bmatrix} + \begin{bmatrix} \Delta x_{n+1} \\ \Delta y_{n+1} \end{bmatrix} = \begin{bmatrix} \alpha \\ \beta \end{bmatrix} + \underbrace{\begin{bmatrix} f_x & f_y \\ g_x & g_y \end{bmatrix}}_{\mathbf{G}}|_{(\alpha, \beta)} \begin{bmatrix} \Delta x_n \\ \Delta y_n \end{bmatrix} + \text{higher order terms}$$

Dropping the higher order terms, we find

$$\begin{bmatrix} \Delta x_{n+1} \\ \Delta y_{n+1} \end{bmatrix} = \mathbf{G} \begin{bmatrix} \Delta x_n \\ \Delta y_n \end{bmatrix} = \mathbf{G}^n \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix}$$

So the convergence of the fixed point iteration depends on what happens to the powers of \mathbf{G} . To get an idea of what we need for convergence, let's recall some linear algebra.

Without loss of generality, let's assume that \mathbf{G} is full rank. Let $\mathbf{G} \in \mathbb{R}^{m \times m}$. (In what we have done so far $m = 2$ but everything extends to other m .) Since \mathbf{G} is full rank, its eigenvectors form a basis for \mathbb{R}^m . Let $\{(\lambda_j, \mathbf{v}_j)\}_{j=1}^m$ denote the eigenpairs. Then for all $\mathbf{x} \in \mathbb{R}^m$, there exists constants $\{\alpha_j\}_{j=1}^m$ such that $\mathbf{x} = \sum_{j=1}^m \alpha_j \mathbf{v}_j$. This means that

$$\begin{aligned} \mathbf{G}\mathbf{x} &= \mathbf{G} \sum_{j=1}^m \alpha_j \mathbf{v}_j \\ &= \sum_{j=1}^m \alpha_j \mathbf{G}\mathbf{v}_j \\ &= \sum_{j=1}^m \alpha_j \lambda_j \mathbf{v}_j \end{aligned}$$

Thus

$$\mathbf{G}^n \mathbf{x} = \sum_{j=1}^m \alpha_j \lambda_j^n \mathbf{v}_j$$

The sum converges when $|\lambda_j| \leq 1$.

Really we want $\mathbf{G}^n \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix} \rightarrow 0$. This means that we need $|\lambda_j| < 1$. We need a stricter condition because we need

$$\begin{bmatrix} \Delta x_{n+1} \\ \Delta y_{n+1} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

In practice, we can look at other norms too.

$$\left\| \begin{bmatrix} \Delta x_{n+1} \\ \Delta y_{n+1} \end{bmatrix} \right\| = \left\| \mathbf{G}^n \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix} \right\| \leq \|\mathbf{G}\|^n \left\| \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix} \right\|$$

It is sufficient for $\|\mathbf{G}\| < 1$. In practice, most people use the l_2 norm.

Theorem 10.6 in the text has a more formal statement of the following.

Theorem 0.1. $\|\mathbf{G}\| \geq \rho(\mathbf{G})$ where $\rho(\mathbf{G})$ denotes the spectral radius defined by $\rho(\mathbf{G}) = \max_j |\lambda_j|$.