

Example problems:

- 1) Numerical calculation of gradient of a function & its comparison with analytical mtd.

lets say. $f: \mathbb{R}^2 \rightarrow \mathbb{R}$.

$$f(x_1, x_2) = 12.069 x_1^2 + 21.504 x_2^2 - 1.7321 x_1 - x_2.$$

We need to find its gradient "numerically".

Step I: Analytical gradient

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{pmatrix} = \begin{pmatrix} 24.138 x_1 - 1.7321 \\ 43.008 x_2 - 1 \end{pmatrix}$$

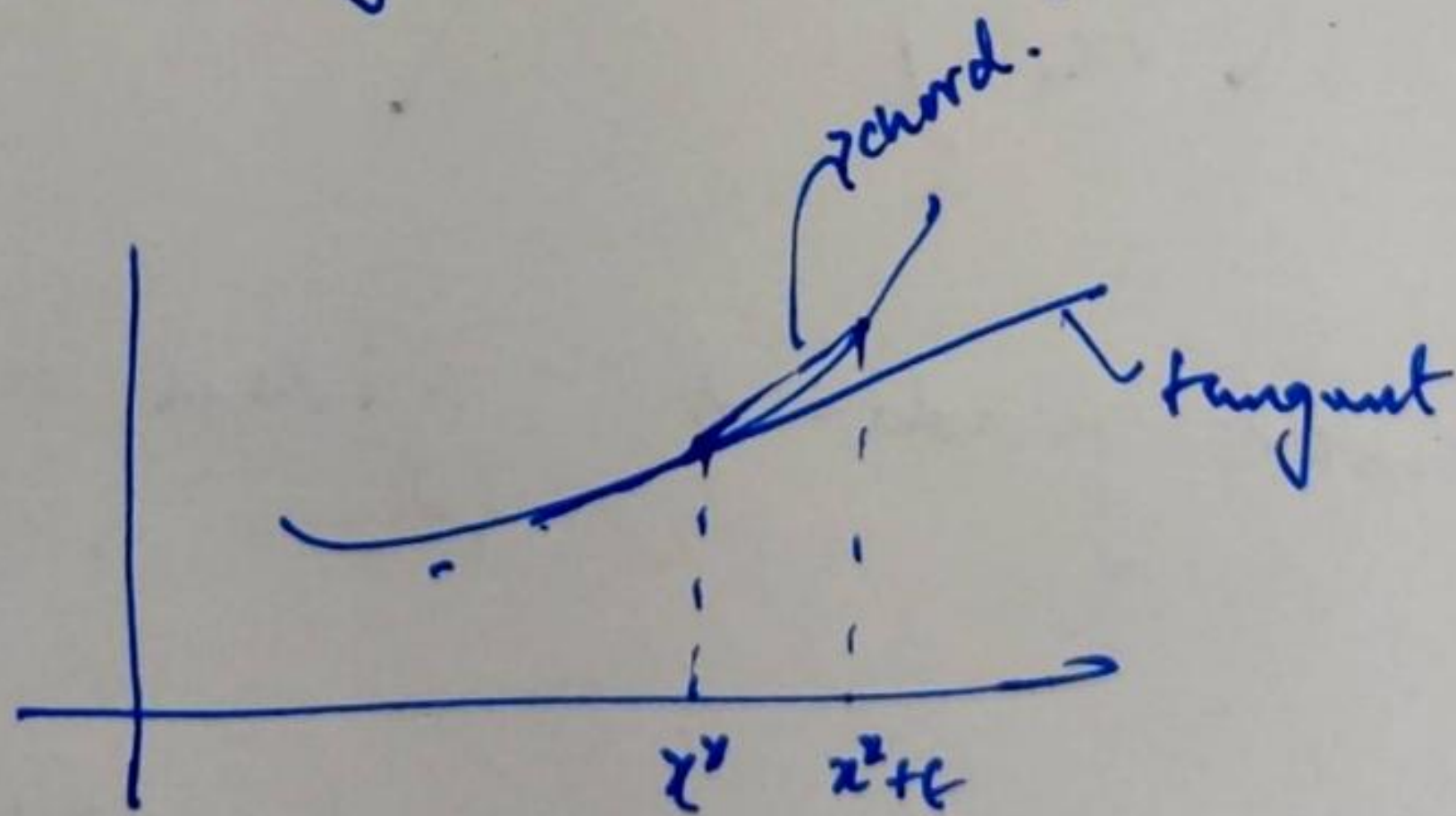
~~Numerical gradient~~ given a pt $\underline{x}_0 = \begin{pmatrix} 5 \\ 6 \end{pmatrix}$

$$\nabla f|_{\underline{x}_0} = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{pmatrix} \Big|_{\underline{x}_0} = \begin{pmatrix} 1.1896 \times 10^2 \\ 2.5705 \times 10^2 \end{pmatrix}$$

$$\underline{x}_0 = \begin{pmatrix} 5 \\ 6 \end{pmatrix}.$$

Numerical gradient calculation:

- Step I: estimate the ϵ (divided difference param)
- each variable is perturbed once with ϵ and the derivatives calculated.
 - Too small a choice of ϵ leads to a very loss of significance error, since two very nearly equal quantities will be subtracted from each other.
 - Small ϵ is also recommended in probs where there is noise in the function evaluation, as occurs when using a nonlinear iterative simulation code to evaluate f .
 - Too large an ϵ results in a large truncation error, graphically it is the error arising due to the mismatch betⁿ the "chord" and "tangent" at the point.



Thumb rule.

$$\epsilon = \max^w (\epsilon_{\min}, 0.01 x_i^0) \text{ works well.}$$

(2)

Step II: A) Using central forward difference

$$\frac{\partial f}{\partial x_i} = \frac{f(x_1^0, x_2^0, \dots, x_i^0 + \epsilon, \dots) - f(x_1^0, x_2^0, \dots, x_i^0, \dots)}{\epsilon}$$

for $\forall i = 1, 2, \dots, n$.

~~or~~

B) Using Backward difference.

$$\frac{\partial f}{\partial x_i} = \frac{f(x_1^0, x_2^0, \dots, x_i^0 - \epsilon, x_{i+1}^0, \dots, x_n^0) - f(x_1^0, x_2^0, \dots, x_i^0, x_{i+1}^0, \dots, x_n^0)}{\epsilon} \quad \forall i = 1, 2, \dots, n$$

C) Using central difference.

$$\frac{\partial f}{\partial x_i} = \frac{f(x_1^0, x_2^0, \dots, x_i^0 + 0.5\epsilon, \dots, x_n^0) - f(x_1^0, x_2^0, \dots, x_i^0 - 0.5\epsilon, \dots, x_n^0)}{\epsilon}$$

Choosing $\epsilon = 0.01 \text{ abs}(x_i^0)$

$\nabla f(\underline{x}^0)$
Forward difference

$$\begin{pmatrix} 1.1956 \times 10^3 \\ 2.5834 \times 10^2 \end{pmatrix}$$

$\nabla f(\underline{x}^0)$
Backward
Central diff

$$\begin{pmatrix} 1.1835 \times 10^2 \\ 2.5576 \times 10^2 \end{pmatrix}$$

$\nabla f(\underline{x}^0)$
Central
Backward diff

$$\begin{pmatrix} 1.1896 \times 10^2 \\ 2.5705 \times 10^2 \end{pmatrix}$$

Problem 2: Numerical evaluation of Hessian matrix.

Assignment: (1)

1) Try to plot error in. (For the prob discussed).

a) Forward diff approximation

b) Backward diff "

c) Central diff "

as you change the value of ϵ , check the convergence. Plot them on the same curve.

2) Take a function.

$$a) f = \frac{4x_2^2 - x_1x_2}{10000(x_2x_1^3 - x_1^4)} \quad @ \quad \underline{x}^0 = \begin{pmatrix} 0.5 \\ 1.5 \end{pmatrix}$$

Find its gradient using Forward diff, Backward & central diff.

Plot the error & ~~converg~~ show convergence graphically.

Clearly show the graph label, title etc.