

DATA SCIENCE: THEORY AND PRACTICE

OPTIMIZATION TECHNIQUES:

Optimization criteria:-

at point \bar{w} , first derivative is zero,
 1st non-zero higher derivative is denoted by n ,

1) $n \rightarrow \text{odd}$; $\bar{w} \rightarrow \text{inflection point}$

2) $n \rightarrow \text{even}$;

a) derivative $\rightarrow +ve$, $\bar{w} \rightarrow \text{local minimum}$

b) derivative $\rightarrow -ve$, $\bar{w} \rightarrow \text{local maximum}$

Aim for optimization:

Finding w for which $J(w)$ is minimum

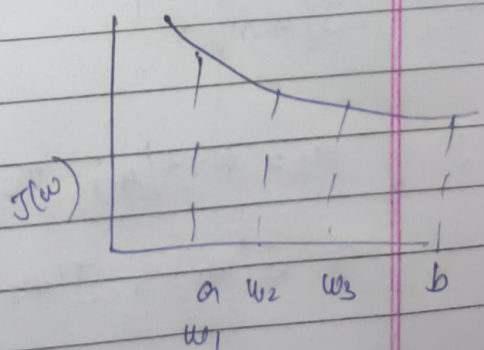
Method 1: Bracketing method
 - Exhaustive search;

$$1: \Delta w = (b - a) / n$$

$$w_1 = a;$$

$$w_2 = w_1 + \Delta w$$

$$w_3 = w_2 + \Delta w$$



$$2: J(w_1) \geq J(w_2) \leq J(w_3)$$

min lies b/w (w_1, w_3)

(or)

$$w_1 = w_2; w_2 = w_3; w_3 = w_2 + \Delta w$$



$$3: w_1 = w_2, w_2 = w_3; w_3 = w_2 + \Delta w$$

If $w_3 \leq b$;
Step 2.

(or) no min between (a, b)

Method 2: Region Elimination method

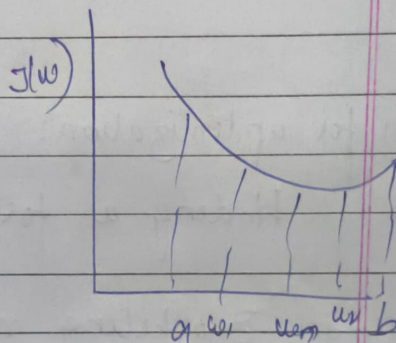
- Interval halving

$$1: a, b$$

$$w_m = a + b/2$$

$$L = b - a$$

$$J(w_m) = ?$$



$$2: w_1 = a + L/4; w_2 = b - L/4$$

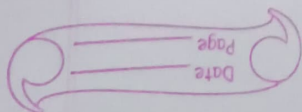
$$J(w_1), J(w_2) = ?$$

$$3: J(w_1) < J(w_m)$$

$$\Rightarrow b = w_m$$

$$\Rightarrow w_m = w_1$$

$$L = b - a$$



if $|L| < \epsilon$,

stop

(or)

$$*) J(w_2) < J(w_m)$$

$$a = w_m; w_m = w_2$$



$$L = b - a$$

if $|L| < \epsilon$

stop

(or)

$$*) a = w_1, b = w_2$$

$$L = b - a$$

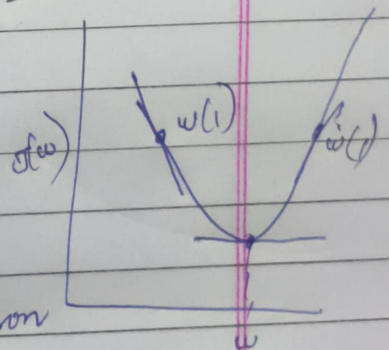
if $|L| < \epsilon$

stop

Method 3: Gradient-based approaches.

Aim: Min \rightarrow point where $J'(w) \approx 0$

$$w^{k+1} = w^k - \frac{J'(w)}{J''(w)}$$



Qn: Interval halving vs Newton-Raphson

Newton-Raphson tends to be faster as the convergence rate is quadratic