

$$\text{Let } M = (\hat{x}, \hat{y})$$

$L_1$  : passes through  $(\cos \theta_1, \sin \theta_1)$  has slope  $\tan \alpha$

$$\therefore y - \sin \theta_1 = \tan \alpha (x - \cos \theta_1) \rightarrow (1)$$

$L_2$  :  $P(\cos \theta_2, \sin \theta_2)$ ;  $m = \tan \beta$

$$y - \sin \theta_2 = \tan \beta (x - \cos \theta_2) \rightarrow (2)$$

Intersection pt. is.

$$\Rightarrow \sin \theta_2 + \tan \beta (x - \cos \theta_2) - \sin \theta_1 = \tan \alpha (x - \cos \theta_1)$$

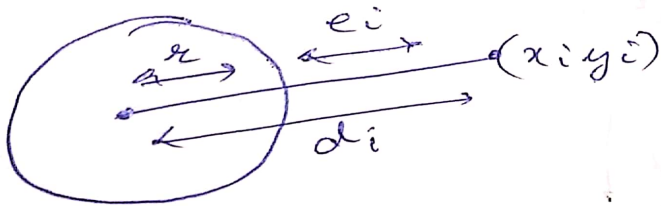
$$\Rightarrow \sin \theta_2 - \sin \theta_1 + x (\tan \beta - \tan \alpha) = \tan \beta \cos \theta_2 - \tan \alpha \cos \theta_1$$

$$\Rightarrow \boxed{\hat{x} = \frac{\tan \beta \cos \theta_2 - \tan \alpha \cos \theta_1 - \sin \theta_2 + \sin \theta_1}{\tan \beta - \tan \alpha}}$$

Substituting in (1)

$$\boxed{\hat{y} = \sin \theta_1 + \tan \alpha (\hat{x} - \cos \theta_1)}$$

$$\text{Let } d_i^2 = \hat{x}_i^2 + \hat{y}_i^2 \text{ for } i = 1, 2, \dots, 5 \text{ data point } (x_i, y_i)$$



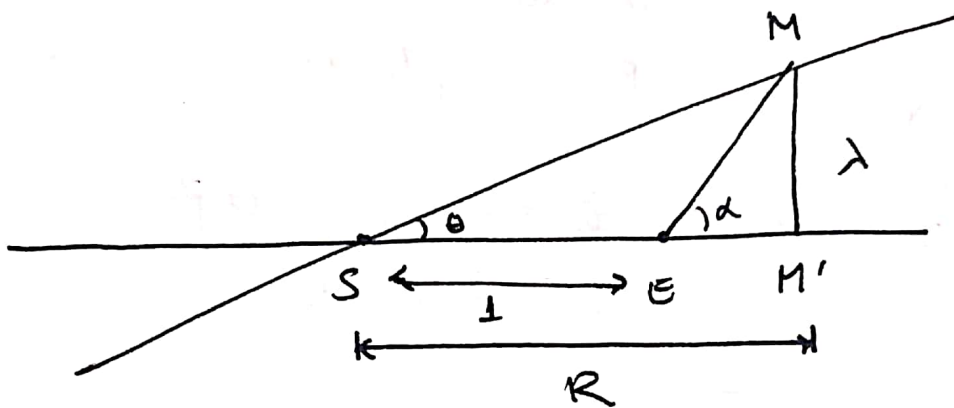
$$L = \sum_{i=1}^r e_i^2$$

$$= \sum_{i=1}^r (d_i - r)^2$$

$$= \sum_{i=1}^r (\sqrt{x_i^2 + y_i^2} - r)^2$$

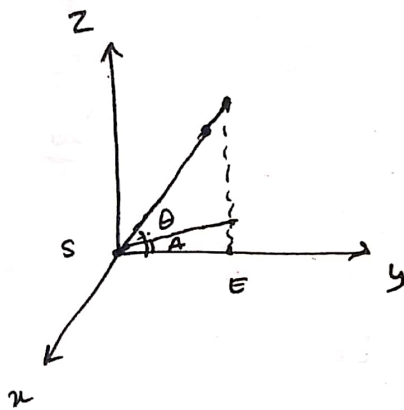
$$L = \sum_{i=1}^r x^2 + x_i^2 + y_i^2 - 2r\sqrt{x_i^2 + y_i^2}$$





$$\tan \alpha = \frac{\lambda}{R-1} \quad ; \quad \tan \theta = \frac{\lambda}{R}$$

$$\theta = \tan^{-1} \left( \frac{1}{R} (R-1) \tan \alpha \right)$$



$\theta = \text{latitude}$   
 $A = \text{longitude}$

$$x = -R \cos \theta \sin A$$

$$z = R \sin \theta$$

$$y = R \cos \theta \cos A$$

Dis of point  $(x_i, y_i, z_i)$  from plane  $d_i = \frac{Ax_i + By_i + Cz_i}{\sqrt{A^2 + B^2 + C^2}}$

$$L = \sum_{i=1}^{12} d_i^2$$

$$L = \frac{\sum v_i}{S}$$

$$\frac{\partial L}{\partial A} = \frac{1}{S} \left( \sum \frac{dv_i}{dA} \right) + \left( \frac{-1}{S^2} \frac{dS}{dA} \right) \sum v_i$$

$$\text{Let } S = A^2 + B^2 + C^2$$

$$\frac{dS}{dA} = 2A, \quad \frac{dS}{dB} = 2B, \quad \frac{dS}{dC} = 2C$$

$$\text{Let } v_i = (Ax_i + By_i + Cz_i)^2$$

$$\frac{dv_i}{dA} = 2v_i x_i$$

$$\frac{dv_i}{dB} = 2v_i y_i$$

$$\frac{dv_i}{dC} = 2v_i z_i$$

$$\frac{\partial l}{\partial A} = \frac{1}{S} \sum (2 v_i x_i) - \frac{1}{S^2} (2A) (\sum v_i)$$

By symmetry

$$\frac{\partial l}{\partial B} = \frac{1}{S} \sum (2 v_i y_i) - \frac{1}{S^2} (2B) (\sum v_i)$$

$$\frac{\partial l}{\partial C} = \frac{1}{S} \sum (2 v_i z_i) - \frac{1}{S^2} (2C) (\sum v_i)$$

3(ii)

Algorithm : Steepest Descent with exact line search using dichotomous search.

Let  $f$  to be the fn to be minimised,

$k=1$ ;  $x_1 \leftarrow \text{Random guess}$   
while  $|\nabla f_k| > \text{tolerance}$

$$\alpha_k^* = \underset{\alpha > 0}{\operatorname{argmin}} f(x_k - \alpha \nabla f_k)$$

where  $\nabla f_k = \nabla f|_{x=x_k}$

$$x_{k+1} = x_k - \alpha_k \nabla f_k$$

$$k = k+1$$

Finding  $\alpha_k$  : Line search using dichotomous search

Input : Interval of uncertainty  $[a, b]$ , tolerance

Init :  $k=0$

$$a_k = a$$

$$b_k = b$$

$$\epsilon (> 0)$$

while  $b_k - a_k > \epsilon$

$$\lambda_k = \frac{a_k + b_k - \epsilon}{2}, \quad \mu_k = \frac{a_k + b_k + \epsilon}{2}$$

if  $f(\lambda_k) > f(\mu_k)$

$$a_{k+1} = \lambda_k$$

$$b_{k+1} = b_k$$

else

$$b_{k+1} = \mu_k$$

$$a_{k+1} = a_k$$

end if

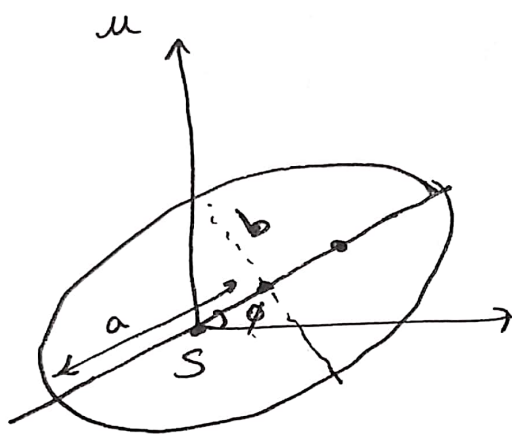
$$k \leftarrow k + 1$$

end while

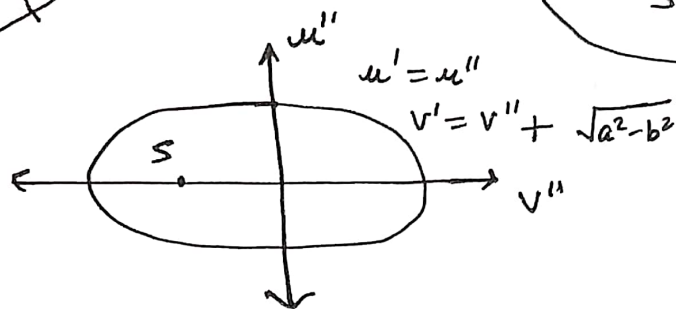
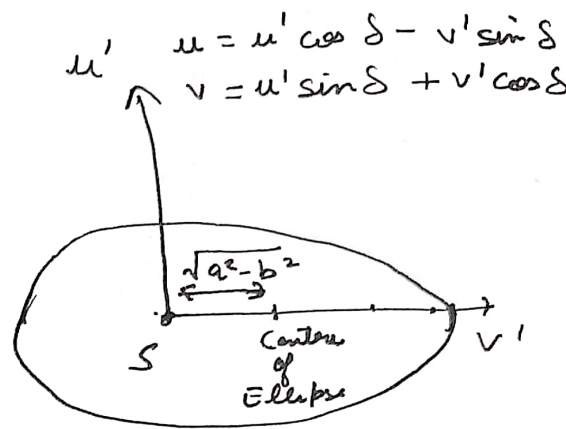
return

$$x^* = \frac{a_k + b_k}{2}$$





Rotation  
by  $\phi = \delta$



Translation  
by  $\sqrt{a^2 - b^2}$

Eq. of ellipse  $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$

$$\Rightarrow \frac{u'^2}{a^2} + \frac{v''^2}{b^2} = 1$$

$$= \frac{u'^2}{a^2} + \frac{(v' + \sqrt{a^2 - b^2})^2}{b^2} = 1$$

$$= \frac{(u \cos \delta + v \sin \delta)^2}{a^2} + \frac{((v \cos \delta - u \sin \delta) + \sqrt{a^2 - b^2})^2}{b^2} = 1$$

Let  $f(u, v; a, b, \delta) = \frac{(u \cos \delta + v \sin \delta)^2}{a^2} + \frac{(v \cos \delta - u \sin \delta + \sqrt{a^2 - b^2})^2}{b^2} - 1$

Then Loss  $\Rightarrow \sum_{(u,v)} f(u, v; a, b, \delta)$

Reason: This loss fn is 0 for perfect fit and has less parameters than number of points.