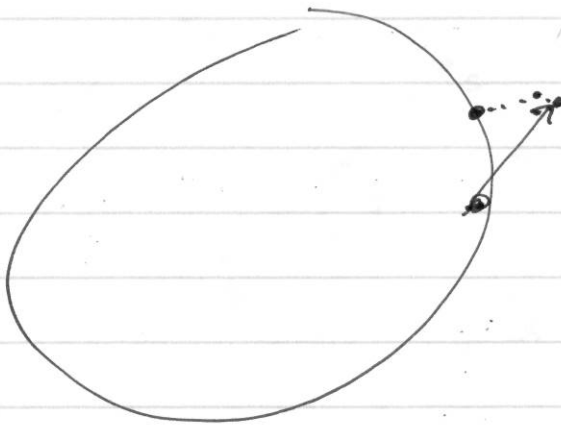


safe screening rule : if it passes, then we are guaranteed to have a 0 coefficient at solution

non-safe rules : may know that w.h.p. have a 0 coefficient at solution  
 SURE OR  
 strong coefficient is small at sol.  
 (means it has little statistical effect)



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$$\|x\|_* = \max_{\|z\| \leq 1} x^T z \quad \text{dual norm}$$

$$\partial \|x\|_* = \text{argmax}_{\|z\| \leq 1} x^T z \quad \text{subgradients.}$$

---


$$\|tz\| = t\|z\|$$

---

if have  $\|x\|_\infty = |x_i|$  ( $i^{\text{th}}$  comp achieved the max)

then  $\text{sign}(x_i) \cdot e_i \in \partial \|x\|_\infty$

$\uparrow$   
 $(0, \dots, 1, \dots, 0)$

$\nwarrow$   $i^{\text{th}}$  spot

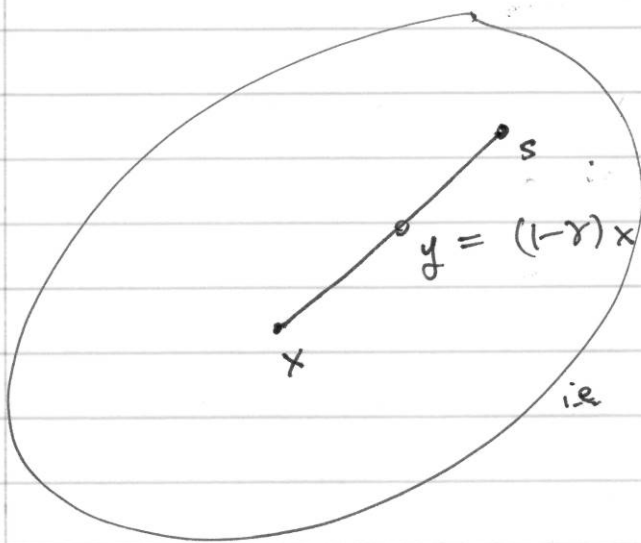
$$f(s) \geq f(x^{(k-1)}) + \nabla f^T (s - x^{(k-1)})$$

min over  $s \in C$  :

$$f^* \geq f(x^{(k-1)}) + \min_{s \in C} \nabla f^T (s - x^{(k-1)})$$

rearrange :

$$- \min_{s \in C} \nabla f^T (s - x^{(k-1)}) \geq f(x^{(k-1)}) - f^*$$



$$f(y) \geq f(x) + \nabla f(x)^T (y - x)$$

ie  $f(y) - f(x) - \nabla f(x)^T (y - x) \geq 0$

