- · Operations that preserve the convexity
  - · non-negative multiples: &f, X>0
  - · Summation: fi+f2
  - · composition of affine function: f(Ax+b)
  - o point-wise maximum:  $f(x) = \max \{f_i(x)\}$ , where  $f_i(x)$  is convex

ex:  $g(x) = \max_{y \in A} f(x,y)$  is convex

\* where is no constraint on A

· partial minimization:

If f(x,y) and C are convex,  $g(x) = \min_{y \in C} f(x,y)$  is convex \* C must be convex

groof>

 $\theta \cdot g(x_1) + (1-\theta) \cdot g(x_2) = \theta \cdot f(x_1, y_1) + (1-\theta) \cdot f(x_2, y_2),$ (where  $y_1 \in \{y \mid \min_{y \in C} g(x_1, y)\}$  and  $y_2 \in \{y \mid \min_{y \in C} g(x_2, y)\}$ )  $f(x_1, y) \stackrel{is}{=} \sum_{convex} g(x_1, y_1) = f(\theta x_1 + (1-\theta)x_2, \theta y_1 + (1-\theta)y_2)$   $c = \sum_{convex} g(x_1, y_1) = f(\theta x_1 + (1-\theta)x_2, y_1)$ 

 $=f(0X_1+(1-0)X_2)$ 

ex: Given  $f(x,y) = [x^T y^T][B^T C][Y]$ ,  $X \in \mathbb{R}^n$ ,  $Y \in \mathbb{R}^n$ ,  $[A B] \in S^{n+m}$ Let  $g(x) = \min_{y \in Y} f(x,y) = X^T (A - BC^T B^T) \times$ , where  $C^T$  is pseudo inverse  $(C^{-1})$ 

- (2) C is convex
- (3) from (1) + (2), g(x) is convex, i.e. A-BCTBT ZO

• Composition  

$$g: \mathbb{R}^n \to \mathbb{R}$$
 and  $h: \mathbb{R} \to \mathbb{R}$   
 $f(x) = h(g(x))$   
 $f'(x) = h'(g(x)) \cdot g'(x)$   
 $f''(x) = h''(g(x)) \cdot [g'(x)]^2 + h'(g(x)) \cdot g''(x)$ 

$$\star$$
 f is convex if  $\begin{cases} 9 \text{ is convex, h is convex, non-decreasing } (h \ge 0) \\ 9 \text{ is concave, h is convex, non-increasing } (h \le 0) \end{cases}$ 

$$*f$$
 is concave if  $\{g \text{ is convex, } h \text{ is concave, non-increasing } (k \leq 0) \}$  is concave, h is concave, non-decreasing  $(k \geq 0)$ 

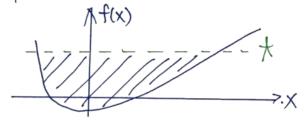
· Different Views of Functions · Equi-potential map



· Graph of f

· Epigraph:

epi 
$$f = \{(x, t) \mid x \in \text{dom } f, f(x) \leq t\}$$



=) f is convex iff epif is convex