1. a) Suppose the first k components are nonzero.

Because  $\sum_{i=1}^{k} x_i = 1$  and  $\log x$  is a concave function,

$$-H(x) \ge \log\left(\frac{\sum_{i=1}^k x_i}{k}\right) = \log\left(\frac{1}{k}\right) = -\log(k).$$

Because  $k \le n$ ,  $H(x) \le \log(k) \le \log(n)$ .

b) 
$$H(\bar{x}) = -\sum_{i=1}^{n} \frac{1}{n} \log \left(\frac{1}{n}\right) = -\log \left(\frac{1}{n}\right) = \log(n).$$

Because  $H(x) \leq \log(n)$ ,  $H(\bar{x})$  sis the maximum of H(x).

$$Hessian(H(x)) = \begin{bmatrix} (-1)^{n-1}(n-2)! \frac{1}{x_1^{n-1}} & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & (-1)^{n-1}(n-2)! \frac{1}{x_n^{n-1}} \end{bmatrix}.$$

Apparently, Hessian(H(x)) is positive-definite.

H(x) is strictly concave.

So  $\bar{x}$  is the unique maximum.

2. a) Let  $\mu = EX$ , for  $a < u < \mu < s < b$ .

Apparently, there exists  $\theta \in [0,1]$  such that  $\theta u + \bar{\theta} s = \mu$ .

Because, 
$$f$$
 is convex. 
$$\frac{f(\mu) - f(s)}{\mu - s} = \frac{f(\theta u + \overline{\theta}s) - f(s)}{\theta u + \overline{\theta}s - s} = \frac{f(\theta u + \overline{\theta}s) - f(s)}{\theta(u - s)} \le \frac{f(u) - f(s)}{(u - s)}$$

For the same reason,  $\frac{f(u)-f(\mu)}{u-\mu} \ge \frac{f(u)-f(s)}{(u-s)}$ .

So, 
$$\frac{f(u)-f(\mu)}{u-\mu} \ge \frac{f(\mu)-f(s)}{\mu-s}$$
.

b) Take 
$$\beta = \sup \frac{f(\mu) - f(s)}{\mu - s}$$
. Because  $\frac{f(u) - f(\mu)}{u - \mu} \ge \frac{f(\mu) - f(s)}{\mu - s}$ ,  $\beta < +\infty$ .

Obviously,  $\beta > -\infty$ .

So, there exists  $\beta$  such that  $f(x) \ge f(\mu) + \beta(x - \mu), \forall x \in (a, b)$ 

c) therefore,  $f(X) \ge f(\mu) + \beta(X - \mu)$ 

Namely,  $f(X) \ge f(EX) + \beta(X - EX)$ .

$$Ef(X) \ge f(EX)$$

$$3. \quad Hessian(\log(1+e^{3x_1+2x_2})) = \begin{bmatrix} \frac{9e^{3x_1+2x_2}}{(1+e^{3x_1+2x_2})^2} & \frac{6e^{3x_1+2x_2}}{(1+e^{3x_1+2x_2})^2} \\ \frac{6e^{3x_1+2x_2}}{(1+e^{3x_1+2x_2})^2} & \frac{4e^{3x_1+2x_2}}{(1+e^{3x_1+2x_2})^2} \end{bmatrix}.$$

$$D_1 = \frac{9e^{3x_1 + 2x_2}}{(1 + e^{3x_1 + 2x_2})^2}, \ D_2 = 0$$

So  $\log (1 + e^{3x_1 + 2x_2})$  is convex.

Therefore,  $\{x: \log(1 + e^{3x_1 + 2x_2}) \le 2\}$  is convex.

Because Ax + b is an affine function, so S is convex.

- 4. a) It's a convex optimization problem.
  - b) It's not a convex optimization problem.