

Algorithmic Game Theory: HW 2

- 1.
- 2.
- 3.
4. Let $f_\epsilon(x) = (1 - \epsilon)^x$ and $g_\epsilon(x) = 1 + \epsilon x$, then

$$\left. \begin{aligned} f_\epsilon(0) &= 1 = g_\epsilon(0) \\ f_\epsilon(1) &= 1 - \epsilon = g_\epsilon(1) \\ f'_\epsilon(x) &= (1 - \epsilon)^x \ln(1 - \epsilon) \\ g'_\epsilon(x) &= \epsilon \end{aligned} \right\} f'_\epsilon(0) = \ln(1 - \epsilon) < 0 = g'_\epsilon(0)$$

also f_ϵ is a convex function as $f''_\epsilon(x) = (1 - \epsilon)^x [\ln(1 - \epsilon)]^2 > 0$ for $\epsilon \in (0, 1/2]$. This this proves $f_\epsilon(x) \leq g_\epsilon(x)$ since the initial gradient of f_ϵ is smaller than g_ϵ