

Appendix: Contract Design for Adaptive Federated Learning

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1 Proof to Lemma 2

- Proof.* 1. The non-negative weighted sum of convex (concave) functions is convex (concave). [?] $g_1(\mathbf{r}) = \sum_{i=1}^I \sum_{j=1}^J N_{i,j} r_{i,j} \gamma_j$ and $g_2(\mathbf{r}) = \sum_{i=1}^I \sum_{j=1}^J N_{i,j} \theta_i r_{i,j}$ are weighted sum of variables $r_{i,j}$, where $N_{i,j}$, γ_j and θ_i are weights. Thus, $g_1(\mathbf{r})$ and $g_2(\mathbf{r})$ are both convex and concave.
2. The scalar composition $f(x) = h(g(x))$ is convex if $g(\cdot)$ is concave, $h(\cdot)$ is convex and \hat{h} is non-increasing, where \hat{h} is the extended-value extension of function h , which assigns the value ∞ to points not in **dom** h . Suppose $h(x) = \frac{1}{\sqrt{x}}$, which is a convex and \hat{h} is non-increasing. Thus, $f(\mathbf{r}) = h(g_1(\mathbf{r}))$ is convex.
3. $C_{com}(\mathbf{r})$ is non-negative weighted sum of convex functions $f(\mathbf{r})$ and $g_2(\mathbf{r})$. Thus, $C_{com}(\mathbf{r})$ is convex. \square