

Figure 1: The loss difference between our two methods and VRDBO. A negative difference indicates that our methods converge faster to the stationary point than VRDBO (The potential reason for the positive difference in the first several steps is that the one-step gradient descent for estimating Hessian-inverse-vector product is not as good as the Neuman series expansion method in the initial stage.). There are 8 workers in this experiment. The training sample's feature x on the k-th worker is generated from a Guassian distribution $\mathcal{N}(\mu_k, \sigma_k)$, where μ_k is generated from a Uniform distribution $\mathcal{U}(-3,3)$ and σ_k is generated from a Uniform distribution $\mathcal{U}(1,25)$.

(b) $LOSS_{DSVRBGD-S} - LOSS_{VRDBO}$