Notes for Lecture 13

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1 Properties of Minimax Optimization

Property 1. In minimax optimization, we have

$$\max_{\mathbf{y} \in \mathcal{Y}} \min_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathbf{y}) \leq \min_{\mathbf{x} \in \mathcal{X}} \max_{\mathbf{y} \in \mathcal{Y}} f(\mathbf{x}, \mathbf{y})$$

Proof. Let

$$\mathbf{y}^* = \underset{\mathbf{y} \in \mathcal{Y}}{\arg \max} \min_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathbf{y})$$
$$\mathbf{x}^* = \underset{\mathbf{x} \in \mathcal{X}}{\arg \min} \max_{\mathbf{y} \in \mathcal{Y}} f(\mathbf{x}, \mathbf{y}),$$

we naturally have

$$\begin{aligned} \max_{\mathbf{y} \in \mathcal{Y}} \min_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathbf{y}) &= \min_{\mathbf{x} \in \mathcal{X}} f(\mathbf{x}, \mathbf{y}^*) \leq f(\mathbf{x}^*, \mathbf{y}^*) \\ \min_{\mathbf{x} \in \mathcal{X}} \max_{\mathbf{y} \in \mathcal{Y}} f(\mathbf{x}, \mathbf{y}) &= \max_{\mathbf{y} \in \mathcal{Y}} f(\mathbf{x}^*, \mathbf{y}) \geq f(\mathbf{x}^*, \mathbf{y}^*), \end{aligned}$$

thus we finish the proof.

2 Convergence Rates of GDA

Theorem 1. If f is l-smooth and convex-concave, let the diameter of \mathcal{X} and \mathcal{Y} be R. Then for fixed t with $\eta = \frac{R}{L\sqrt{t}}$, where $L = 2lR + \|\nabla f(\mathbf{x}_0, \mathbf{y}_0)\|_2$, we have

$$\max_{\mathbf{y} \in \mathcal{Y}} f\left(\frac{1}{t} \sum_{k=1}^{t} \mathbf{x}_{k}, \mathbf{y}\right) - \min_{\mathbf{x} \in \mathcal{X}} f\left(\mathbf{x}, \frac{1}{t} \sum_{k=1}^{t} \mathbf{y}_{k}\right) \leq \frac{2LR}{\sqrt{t}}.$$

Proof. It follows that

$$f(\mathbf{x}_{k}, \mathbf{y}) - f(\mathbf{x}, \mathbf{y}_{k}) = f(\mathbf{x}_{k}, \mathbf{y}) - f(\mathbf{x}_{k}, \mathbf{y}_{k}) + f(\mathbf{x}_{k}, \mathbf{y}_{k}) - f(\mathbf{x}, \mathbf{y}_{k})$$

$$\leq \nabla_{\mathbf{y}} f(\mathbf{x}_{k}, \mathbf{y}_{k})^{\top} (\mathbf{y} - \mathbf{y}_{k}) + \nabla_{\mathbf{x}} f(\mathbf{x}_{k}, \mathbf{y}_{k})^{\top} (\mathbf{x}_{k} - \mathbf{x})$$

$$= \frac{1}{\eta} (\tilde{\mathbf{y}}_{k+1} - \mathbf{y}_{k})^{\top} (\mathbf{y} - \mathbf{y}_{k}) + \frac{1}{\eta} (\mathbf{x}_{k} - \tilde{\mathbf{x}}_{k+1})^{\top} (\mathbf{x}_{k} - \mathbf{x})$$

$$= \frac{1}{2\eta} [\|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}_{k}\|_{2}^{2} + \|\mathbf{y}_{k} - \mathbf{y}\|_{2}^{2} - \|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}\|_{2}^{2}] +$$

$$\frac{1}{2\eta} [\|\mathbf{x}_{k} - \tilde{\mathbf{x}}_{k+1}\|_{2}^{2} + \|\mathbf{x} - \mathbf{x}_{k}\|_{2}^{2} - \|\mathbf{x} - \tilde{\mathbf{x}}_{k+1}\|_{2}^{2}]$$

$$\leq \frac{1}{2\eta} [\eta^{2}L^{2} + \|\mathbf{y}_{k} - \mathbf{y}\|_{2}^{2} - \|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}\|_{2}^{2}] +$$

$$\frac{1}{2\eta} [\eta^{2}L^{2} + \|\mathbf{x} - \mathbf{x}_{k}\|_{2}^{2} - \|\mathbf{x} - \tilde{\mathbf{x}}_{k+1}\|_{2}^{2}].$$

With the property of projection

$$\|\mathbf{x} - \mathcal{P}_{\mathcal{C}}(\mathbf{x})\|_{2}^{2} + \|\mathbf{z} - \mathcal{P}_{\mathcal{C}}(\mathbf{x})\|_{2}^{2} \le \|\mathbf{x} - \mathbf{z}\|_{2}^{2}, \quad \mathbf{z} \in \mathcal{C}$$

we have

$$\|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}_{k+1}\|_{2}^{2} + \|\mathbf{y} - \mathbf{y}_{k+1}\|_{2}^{2} \le \|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}\|_{2}^{2}$$
$$\|\mathbf{y} - \mathbf{y}_{k+1}\|_{2}^{2} \le \|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}\|_{2}^{2},$$

then we obtain

$$f(\mathbf{x}_{k}, \mathbf{y}) - f(\mathbf{x}, \mathbf{y}_{k}) \leq \frac{1}{2\eta} \left[\eta^{2} L^{2} + \|\mathbf{y}_{k} - \mathbf{y}\|_{2}^{2} - \|\tilde{\mathbf{y}}_{k+1} - \mathbf{y}\|_{2}^{2} \right] + \frac{1}{2\eta} \left[\eta^{2} L^{2} + \|\mathbf{x} - \mathbf{x}_{k}\|_{2}^{2} - \|\mathbf{x} - \tilde{\mathbf{x}}_{k+1}\|_{2}^{2} \right]$$

$$\leq \frac{1}{2\eta} \left[\eta^{2} L^{2} + \|\mathbf{y}_{k} - \mathbf{y}\|_{2}^{2} - \|\mathbf{y}_{k+1} - \mathbf{y}\|_{2}^{2} \right] + \frac{1}{2\eta} \left[\eta^{2} L^{2} + \|\mathbf{x} - \mathbf{x}_{k}\|_{2}^{2} - \|\mathbf{x} - \mathbf{x}_{k+1}\|_{2}^{2} \right],$$

after apply it recursively, we obtain

$$\frac{1}{t} \sum_{k=1}^{t} \left(f(\mathbf{x}_{k}, \mathbf{y}) - f(\mathbf{x}, \mathbf{y}_{k}) \right) \leq \frac{1}{2\eta} \left[\eta^{2} L^{2} + \frac{\|\mathbf{y}_{1} - \mathbf{y}\|_{2}^{2} - \|\mathbf{y}_{t+1} - \mathbf{y}\|_{2}^{2}}{t} \right] + \frac{1}{2\eta} \left[\eta^{2} L^{2} + \frac{\|\mathbf{x} - \mathbf{x}_{1}\|_{2}^{2} - \|\mathbf{x} - \mathbf{x}_{t+1}\|_{2}^{2}}{t} \right] \\
\leq \frac{1}{2\eta} \left[\eta^{2} L^{2} + \frac{\|\mathbf{y}_{1} - \mathbf{y}\|_{2}^{2}}{t} \right] + \frac{1}{2\eta} \left[\eta^{2} L^{2} + \frac{\|\mathbf{x} - \mathbf{x}_{1}\|_{2}^{2}}{t} \right] \\
\leq \frac{1}{2\eta} \left(\eta^{2} L^{2} + \frac{R^{2}}{t} \right) + \frac{1}{2\eta} \left(\eta^{2} L^{2} + \frac{R^{2}}{t} \right) \\
\leq \eta L^{2} + \frac{R^{2}}{t\eta} \\
\leq \frac{2LR}{\sqrt{t}}.$$