

A Deterministic Streaming Sketch for Ridge Regression

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Ridge Regression

Given : $\mathbf{A} \in \mathbb{R}^{n \times d}$, $\mathbf{b} \in \mathbb{R}^n$ in rows.

Goal : $\mathbf{x}_\gamma \equiv \operatorname{argmin}_{\mathbf{x} \in \mathbb{R}^d} (\|\mathbf{Ax} - \mathbf{b}\|^2 + \gamma \|\mathbf{x}\|^2)$

$$= \frac{(\mathbf{A}^\top \mathbf{A} + \gamma \mathbf{I})^{-1} \mathbf{A}^\top \mathbf{b}}{\quad \quad \quad}$$

Space: $O(d^2)$ $O(d)$

Time: $O(d^3)$ $O(nd)$

Our approach: Use Frequent Directions (FD) (Liberty, 2013) to estimate $\mathbf{A}^\top \mathbf{A}$ in stream.

Frequent Directions Ridge Regression

Algorithm FDRR (Based on Frequent Directions)

Input: $\mathbf{A} \in \mathbb{R}^{n \times d}$, $\mathbf{b} \in \mathbb{R}^n$, ℓ, γ

$\Sigma \leftarrow \mathbf{0}^{\ell \times \ell}$, $\mathbf{V}^\top \leftarrow \mathbf{0}^{\ell \times d}$, $\mathbf{c} \leftarrow \mathbf{0}^d$

$\mathbf{C} = \Sigma \mathbf{V}^\top$

for batch $\mathbf{A}_\ell \in \mathbf{A}$, $\mathbf{b}_\ell \in \mathbf{b}$ **do**

$\Sigma', \mathbf{V}'^\top \leftarrow \text{svd} \left(\left[\mathbf{C}^\top; \mathbf{A}_\ell^\top \right]^\top \right)$

$\Sigma \leftarrow \sqrt{\Sigma_\ell'^2 - \sigma_{\ell+1}^2 \mathbf{I}_\ell}$

$\mathbf{V} \leftarrow \mathbf{V}'_\ell$

$\mathbf{C} = \Sigma \mathbf{V}^\top$

$\mathbf{c} \leftarrow \mathbf{c} + \mathbf{A}_\ell^\top \mathbf{b}_\ell$

end for

$\mathbf{c}' = \mathbf{V}^\top \mathbf{c}$

$\hat{\mathbf{x}}_\gamma \leftarrow \mathbf{V}(\Sigma^2 + \gamma \mathbf{I})^{-1} \mathbf{c}' + \gamma^{-1}(\mathbf{c} - \mathbf{V} \mathbf{c}')$

return $\mathbf{C} \hat{\mathbf{x}}_\gamma$

} Initialization

} size ℓ batch \mathbf{A}_ℓ and \mathbf{b}_ℓ

} Frequent Directions

} Compute $\mathbf{A}^\top \mathbf{b}$ on the fly

} Return the solution $\hat{\mathbf{x}}_\gamma = (\mathbf{C}^\top \mathbf{C} + \gamma \mathbf{I})^{-1} \mathbf{A}^\top \mathbf{b} = (\mathbf{V} \Sigma^2 \mathbf{V}^\top + \gamma \mathbf{I})^{-1} \mathbf{c}$

} Recall the RR solution $\mathbf{x}_\gamma = (\mathbf{A}^\top \mathbf{A} + \gamma \mathbf{I})^{-1} \mathbf{A}^\top \mathbf{b}$

Frequent Directions Ridge Regression

Running time: $O(nd\ell)$, required space: $O(d\ell)$. Note that $\ell \leq d$.

If

$$\ell \geq \frac{\|\mathbf{A} - \mathbf{A}_k\|_F^2}{\varepsilon\gamma} + k, \quad \text{or} \quad \gamma \geq \frac{\|\mathbf{A} - \mathbf{A}_k\|_F^2}{\varepsilon(\ell - k)}$$

Then

- $\|\hat{\mathbf{x}}_\gamma - \mathbf{x}_\gamma\| \leq \varepsilon \|\mathbf{x}_\gamma\|$, or the coefficient error $\frac{\|\hat{\mathbf{x}}_\gamma - \mathbf{x}_\gamma\|}{\|\mathbf{x}_\gamma\|} \leq \varepsilon$
- $|\hat{\mathbf{x}}_\gamma^\top \mathbf{a} - \mathbf{x}_\gamma^\top \mathbf{a}| \leq \varepsilon \|\mathbf{x}_\gamma\| \|\mathbf{a}\|$ for any $\mathbf{a} \in \mathbb{R}^d$
- $\mathcal{B}^2(\hat{\mathbf{x}}_\gamma) \leq \left(1 + \frac{\varepsilon^2}{\gamma^2} \|\mathbf{A}\|_2^4\right) \mathcal{B}^2(\mathbf{x}_\gamma)$
- $\mathcal{V}(\hat{\mathbf{x}}_\gamma) \leq \left(1 + \frac{1}{\gamma} \|\mathbf{A}\|_2^2\right) \mathcal{V}(\mathbf{x}_\gamma)$

Experiments

