

[HETEROGENEOUS BIG DATA]
GRADIENT

$$a_y' = \frac{1}{n} \sum_{i=0}^{n-1} y[i] \rightarrow \gamma a_x, \quad a_x' = \frac{1}{L} \sum_{i=0}^{L-1} x[i]$$

let: $\Pi_1 = \frac{1}{L} \gamma^T X^T \mathbb{1}_L$ $\nabla_x \Pi_1 = \frac{1}{L} \mathbb{1}_L \gamma^T$
 $\left(\frac{1}{L} \mathbb{1}_L^T X \right)^T = \frac{1}{L} \langle \gamma, X^T \mathbb{1}_L \rangle$ $\nabla_y \Pi_1 = \frac{1}{L} X^T \mathbb{1}_L = \text{mean of } X$

ALL INDEXING
IS ZERO PADDED
OUT OF BOUNDS

$$a_y^2[l] = \frac{1}{n} \sum_{i=0}^{n-1} y[i] y[i+l] \rightarrow \sum_{k=1}^K \gamma_k (a_{x_k}^2[l]) + \text{bias}$$

$0 \leq l \leq L-1$ $\gamma_k = \frac{L m_k}{n}$ $m_k = \# \text{ occurrences of } x_k \text{ in } y$

$$= \frac{1}{L} \sum_{i=0}^{L-1} x_k[i] x_k[i+l]$$

bias: if $l=0$: σ^2 in total: we could ignore that entry, or estimate σ .

$$\Pi_2[l] = \gamma^T a^2 + \sigma^2 \delta_0$$

$$a^2 = \begin{bmatrix} \end{bmatrix} \in \mathbb{R}^K$$

$$a^2[k] = \frac{1}{L} \langle x_k[I_1], x_k[I_2] \rangle$$

$$= \frac{1}{L} \langle \mathbb{I}_1 x_k, \mathbb{I}_2 x_k \rangle$$

$$\frac{\partial}{\partial x_k} (\Pi_2[l]) = \gamma_k \frac{1}{L} \langle \mathbb{I}_1^T \mathbb{I}_2 + \mathbb{I}_2^T \mathbb{I}_1 \rangle x_k$$