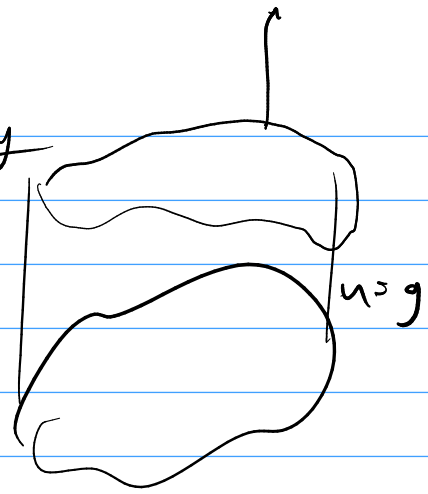


Today



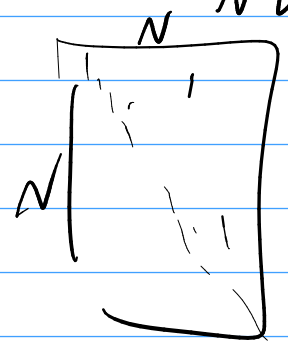
$$\Delta u = 0$$

$$N \times N$$

$$k \ll N$$

$$N^\alpha k^\beta$$

right now: $N^2 k$
 $N k^2$



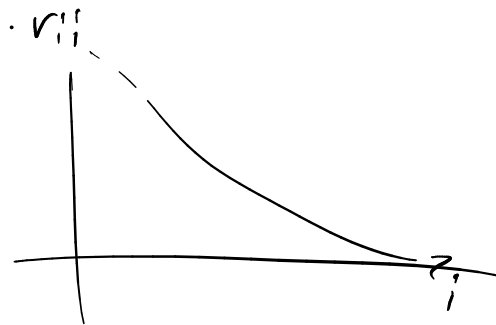
- R R Q R
- 1 D

Rank-revealing/pivoted QR

Sometimes the SVD is too *good* (aka expensive)—we may need less accuracy/weaker promises, for a significant decrease in cost.

$$\begin{aligned} A \Pi &= QR \\ &= Q \left(\begin{array}{c|c} R_{11} & R_{12} \\ \hline & R_{22} \end{array} \right) \end{aligned}$$

$$\sigma_{k+1} \in \|R_{22}\|_2$$



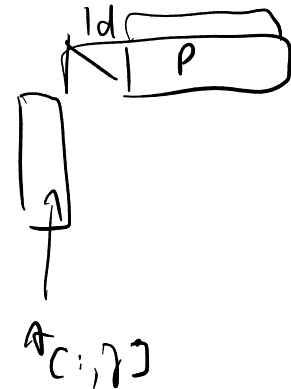
Interpolative Decomposition (ID)

Sometimes it would be helpful to know *which columns of A* contribute 'the most' to the rank.

(rather than have the waters muddied by an orthogonal transformation like in QR)

$$A \approx A_{[:,j]} P$$

$$j \in \mathbb{N}^k$$



$$\text{Set } B = QR_{11}$$

$$\begin{matrix} \text{RRQR} \\ \hookrightarrow A = Q(R_{11} R_{12}) \end{matrix}$$

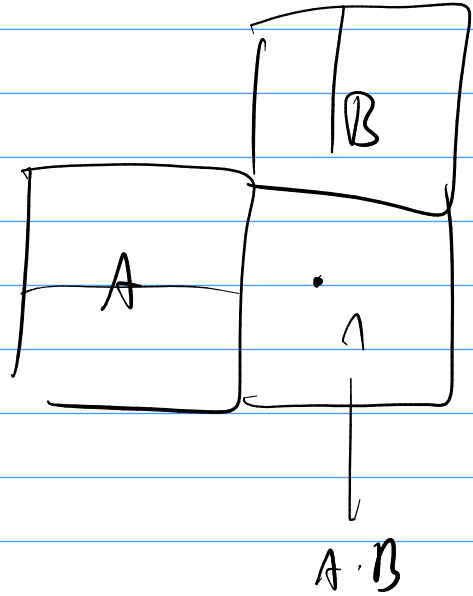
$$P = \left(\underbrace{\text{Id}}_k \quad R_{11}^{-1} R_{12} \right)$$

$$\begin{aligned} BP &= QR_{11} \left(\text{Id} \quad R_{11}^{-1} R_{12} \right) \\ &= Q \left(R_{11} \quad R_{12} \right) \approx A \end{aligned}$$

\hookrightarrow

- P is well-cond
- entries $| \cdot | \leq 2$.

$A \cdot B$



What does the ID buy us?

Specifically: Name a property that the ID has that other factorizations do not have.

$$A \approx Q Q^T A$$

column ID

$$A = A_{[:,j]} P$$

Row ID:

$$Q = P Q_{[j,:]}$$

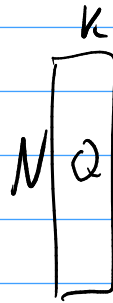
$$A_{[j,:]} \approx [Q Q^T A]_{[j,:]}$$

$$\approx Q_{[j,:]} Q^T A$$

$$= [P Q_{[j,:]}]_{[j,:]} Q^T A$$

$$= \begin{matrix} \text{id} \\ \text{P}_{(j,i)} \end{matrix} Q_{(j,i)} Q^T A$$

$$= Q_{(j,i)} Q^T A$$
$$\approx A$$



ID Q vs ID A

What does row selection mean for the LRA?

Leveraging the ID

Build a low-rank SVD with row extraction.

1. Obtain row subset J and upsampler P .
2. Compute a row QR

$$\underbrace{(A_{[J,:]})^T}_{N \times k} = \underbrace{\bar{Q}}_{N \times k} \underbrace{\bar{R}}_{k \times k}$$

3. Upsample row coefficients: \bar{R}^T

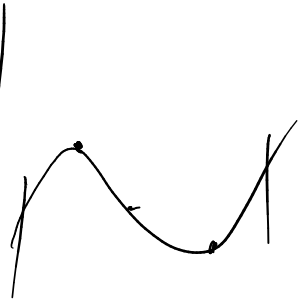
$$Z = P \bar{R}^T$$

$N \times k$ $N \times k$ $k \times k$

4. SVD $Z = U \Sigma V^T$

$$N \begin{array}{|c} k \end{array}$$

$$N^2 k \in \text{bad}$$
$$N k^2$$



$$\begin{aligned}U &\in (\tilde{Q} \tilde{V})^T \\&= \underbrace{U \tilde{V}^T}_{\tilde{Z}} \tilde{Q}^T \\&= \tilde{Z} \tilde{Q}^T \\&= P \tilde{U}^T \tilde{Q}^T \\&= P A_{(j_i)} \\&\approx A\end{aligned}$$

Where are we now?

3 Rank and Smoothness

Punchline

What do (numerical) rank and smoothness have to do with each other?

Recap: Multivariate Taylor

How does Taylor's theorem get generalized to multiple dimensions?