LARSON—MATH 511—CLASSROOM WORKSHEET 24 Low-Rank and Compressed Sensing

Sage/CoCalc

- 1. (a) Start the Chrome browser.
 - (b) Go to http://cocalc.com
 - (c) Login (likely using your VCU email address).
 - (d) You should see an existing Project for our class. Click on that.
 - (e) Click "New", then "Sage Worksheet", then call it c24.
- 2. (**Rank-k changes**). Let $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$. Let $UV^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$ (rank-2).

How does subtraction of UV^T change the inverse of A? (We know the inverse of A. What is the inverse of $M = A - UV^T$?)

3. (Sherman-Morrison-Woodbury formula) We checked the formula:

$$M^{-1} = A^{-1} + A^{-1}U(I - V^{T}A^{-1}U)^{-1}V^{T}A^{-1}$$

with $M = A - UV^T$ and U, V^T from the previous example.

Compute M^{-1} using this formula and compare it with the previous computation.

(Vandermonde Matrices) We showed: Given n+1 points you can find a unique degree-n polynomial that fits them; that is, given points $(x_0, y_0), (x_1, y_1), \ldots, (x_n, y_n)$, you can find a unique function $f(x) = c_0 + c_1 x^1 + \ldots + c_n x^n$ such that $f(x_0) = y_0, f(x_1) = y_1, \ldots, f(x_n) = y_n$ (assuming all x_i 's are different of course).

Find it by solving
$$V\hat{c} = \hat{y}$$
, where $\hat{c} = \begin{bmatrix} c_0 \\ c_1 \\ \dots \\ c_n \end{bmatrix}$, $\hat{y} = \begin{bmatrix} y_0 \\ y_1 \\ \dots \\ y_n \end{bmatrix}$, and

$$V = \begin{bmatrix} 1 & x_0^1 & x_0^2 & \dots & x_0^n \\ 1 & x_1^1 & x_1^2 & \dots & x_1^n \\ \dots & \dots & \dots & \dots \\ 1 & x_n^1 & x_n^2 & \dots & x_n^n \end{bmatrix}$$

We proved: V is invertible iff the x_i s are all distinct.

4. (**Philosophical Implications**). What is the next term in the sequence $1, 2, 3, 4 \dots$? Choose any number for the next term. Find the Vandermonde matrix V for your sequence, find $V^{-1}\hat{y}$ where \hat{y} contains the sequence entries, form a degree-4 polynomial that fits these sequence terms.

Return to Least Squares

(Sherman-Morrison-Woodbury formula) If $M = A - UV^T$ and U, V^T then:

$$M^{-1} = A^{-1} + A^{-1}U(I - V^{T}A^{-1}U)^{-1}V^{T}A^{-1}$$

Suppose we have m data points (each with k features):

$$(A_{1,1}, A_{1,2}, \dots, A_{1,k}, b_1)$$

 $(A_{2,1}, A_{2,2}, \dots, A_{2,k}, b_2)$
 \dots
 $(A_{m,1}, A_{m,2}, \dots, A_{m,k}, b_m)$

and we have already found the least-squares solution \hat{c} to $f(x) = c_1x_1 + c_2x_2 + \ldots + c_kx_k$, where:

$$f(A_{i,1}, A_{i,2}, \dots, A_{i,k}) = b_i$$

for i = 1, 2, ..., m, where $\hat{c} = (A^T A)^{-1} A^T \hat{b}$.

Suppose we then get a new data point:

$$(A_{m+1,1}, A_{m+1,2}, \dots, A_{m+1,k}, b_{m+1})$$

How can we use the Sherman-Morrison-Woodbury formula to efficiently update $(A^TA)^{-1}$ (for the *updated* data matrix A, given that we already know the $(A^TA)^{-1}$ from the original data points)?

The Derivative of A^{-1}

5. Let
$$A(t) = \begin{bmatrix} t & \frac{1}{t} \\ t^2 & t^2 + 1 \end{bmatrix}$$
. Find $A(1)$, $A(2)$.

- 6. Find $\frac{dA}{dt}$.
- 7. Let A = A(1) and B = A(2). Are they invertible?
- 8. Let $\Delta A = B A$. Find ΔA .
- 9. (A Very Useful Formula). Check: $B^{-1} A^{-1} = B^{-1}(A B)A^{-1}$.
- 10. Use this to find $\frac{\Delta A^{-1}}{\Delta t}$ and $\frac{dA^{-1}}{dt}$.

Getting your classwork recorded

When you are done, before you leave class...

- 1. Click the "Make pdf" (Adobe symbol) icon and make a pdf of this worksheet. (If CoCalc hangs, click the printer icon, then "Open", then print or make a pdf using your browser).
- 2. Send me an email with an informative header like "Math 511—c24 worksheet attached" (so that it will be properly recorded).