

**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  $(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)$ , you can find a unique function  $f(x) = c_0 + c_1x^1 + \dots + c_nx^n$  such that  $f(x_0) = y_0, f(x_1) = y_1, \dots, f(x_n) = y_n$  (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

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

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...? 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):

$$(A_{0,0}, A_{0,1}, \dots, A_{0,k-1}, b_0)$$

$$(A_{1,0}, A_{1,1}, \dots, A_{1,k-1}, b_1)$$

...

$$(A_{m-1,0}, A_{m-1,1}, \dots, A_{m-1,k-1}, b_{m-1})$$

and we have *already* found the least-squares solution  $\hat{c}$  to  $f(x) = c_0x_0 + c_1x_1 + \dots + c_{k-1}x_{k-1}$ , where:

$$f(A_{i,0}, A_{i,1}, \dots, A_{i,k-1}) = b_i$$

for  $i = 0, 1, \dots, m-1$ , where  $A^T A \hat{c} = (A^T A)^{-1} A^T \hat{b}$ .

Suppose we then get a new data point:

$$(A_{m,0}, A_{m,1}, \dots, A_{m,k-1}, b_m)$$

How can we use the Sherman-Morrison-Woodbury formula to efficiently update  $(A^T A)^{-1}$  (for the *updated* data matrix  $A$ , given that we already know the  $(A^T A)^{-1}$  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  $\Delta A$ .
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).