

The 2 by 2 Real Pseudo Inverse

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This is a brief note on the math behind the direct PRESS statistic calculation¹ found in the RcppDynProg package².

The actual ‘C++’ code³ is a bit ugly and intimidating. That is because we are using a verbose scalar notation to represent matrix concepts. In matrix notation we are solving a linear system by inverting a two by two matrix.⁴ We are in turn inverting the two by two matrix by exploiting the following well know rule of how to invert a two by two matrix.

If $ad - bc$ is not zero then:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Throughout a , b , c , and d all real scalars.

This can be re-written as the following general relation.

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} = (ad - bc) \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

The above can be directly checked just by applying the rules for multiplying matrices to the left two matrices. This has a particularly pleasant presentation if we recognize that $ad - bc$ is the determinant of the left matrix and use the traditional vertical bar determinant notation.

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Now there is an issue of what to do when $ad - bc$ is zero. For our implementation we apply Tikhonov regularization⁵ which (barring the exact numeric coincidence of minus two times the expected value of the independent variable equaling our regularization constant) is going to be non-singular. For our actual

¹<http://www.win-vector.com/blog/2014/09/estimating-generalization-error-with-the-press-statistic/>

²<https://github.com/WinVector/RcppDynProg>

³https://github.com/WinVector/RcppDynProg/blob/master/src/xlin_fits.cpp

⁴Yes, there are the usual admonitions that one should not invert a matrix to solve a linear system, but for systems this small they do not apply.

⁵https://en.wikipedia.org/wiki/Tikhonov_regularization

application, we could simply switch degenerate situations to the out-of sample mean implementation⁶.

But, for fun, let's play with the math a bit.

There is an additional lesser known algebraic relation for two by two matrices.

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} a & c \\ b & d \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = (a^2 + b^2 + c^2 + d^2) \begin{pmatrix} a & b \\ c & d \end{pmatrix} + (ad - bc) \begin{pmatrix} -d & c \\ b & -a \end{pmatrix}$$

Or (using transpose, matrix squared Frobenius norm, and determinant notation):

Theorem 1 For any real 2 by 2 matrix $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$ we have:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix}^{\top} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \left\| \begin{pmatrix} a & b \\ c & d \end{pmatrix} \right\|_2^2 \begin{pmatrix} a & b \\ c & d \end{pmatrix} + \left| \begin{pmatrix} a & b \\ c & d \end{pmatrix} \right| \begin{pmatrix} -d & c \\ b & -a \end{pmatrix}$$

. The superscript "top" denoting the transpose operation, the $\|\cdot\|_2^2$ denoting sum of squares norm, and the single $|\cdot|$ denoting determinant.

□

This means, if $ad - bc$ is zero then:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix}^{\top} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \left\| \begin{pmatrix} a & b \\ c & d \end{pmatrix} \right\|_2^2 \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

Once we [confirm the above relation⁷ we can also confirm that if $a^2 + b^2 + c^2 + d^2$ is not zero, then the following matrix:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^+ = \frac{1}{a^2 + b^2 + c^2 + d^2} \begin{pmatrix} a & c \\ b & d \end{pmatrix}$$

satisfies all of the conditions for being the Moore-Penrose inverse⁸ (or pseudo-inverse) of our original matrix. The superscript-plus denoting the Moore-Penrose inverse operation.

Or in transpose notation:

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^+ = \frac{1}{a^2 + b^2 + c^2 + d^2} \begin{pmatrix} a & b \\ c & d \end{pmatrix}^{\top}$$

This is called the pseudo-inverse because it acts like an inverse, even for non-invertible matrices. For A^+ to be a Moore-Penrose inverse we must confirm it obeys the following relations:

⁶https://github.com/WinVector/RcppDynProg/blob/master/src/const_costs.cpp

⁷<https://github.com/WinVector/RcppDynProg/blob/master/extras/PseudoInverse.ipynb>

ipyb

⁸https://en.wikipedia.org/wiki/MoorePenrose_inverse

$$\begin{aligned}
AA^+A &= A \\
A^+AA^+ &= A^+ \\
(AA^+)^\top &= AA^+ \\
(A^+A)^\top &= A^+A
\end{aligned}$$

Theorem 1 lets us check the first relation, the other follow quickly as our A^+ is a simple scalar multiple of the transpose.⁹

All of the above check relations would be true for a classic inverse. We can think of A^+ as almost canceling a single A to the left or the A to the right.

Theorem 2 For any real 2 by 2 matrix $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$

The Moore-Penrose inverse is:

- When $ad - bc$ is not zero:

$$\frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

- When $ad - bc$ is zero and $a^2 + b^2 + c^2 + d^2$ is not zero:

$$\frac{1}{a^2 + b^2 + c^2 + d^2} \begin{pmatrix} a & c \\ b & d \end{pmatrix}$$

- Otherwise:

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

□

For general matrices the situation is much more complicated. The wealth of symmetries and relations is really kind of neat.

⁹With the same scaling for both A^+ and $(A^\top)^+$.