# Methods in Transportation Econometrics and Statistics (Master) 

## Winter semester 2023/24, Solutions to Tutorial No. 2

## Solution to Problem 2.1: Matrix Rules

(a) For scalar products, a simple evaluation shows directly

$$
\vec{a}^{\prime} \vec{b}=\vec{b}^{\prime} \vec{a}=\sum_{i} a_{i} b_{i} .
$$

Notice the swapping of transpositions with the swapping of the position. In fact, this is a special case of a rule of transpositions on products to be discussed further below.

For general matrix products, commutativity does not apply, at least not without additional transposition: If $\underline{\underline{A}}$ is a $n \times m$ matrix and $B$ a $m \times k$ matrix, $\underline{\underline{A B}}$ is a $n \times k$ matrix while $\underline{\underline{B A}}$ is not even defined for $n \neq m$ or $m \neq k$. For square matrices $n=m=k$, we can check the 11-component:

$$
(\underline{\underline{A B}})_{11}=\sum_{k} a_{1 k} b_{k 1}, \quad(\underline{\underline{B A}})_{11}=\sum_{k} b_{1 k} a_{k 1} .
$$

In the first product, we need the matrix elements $a_{11}, a_{12}, \ldots$ while we need $a_{11}, a_{21}, \ldots$ for the second. This is not equivalent in case of non-symmetric matrices. Comparing the 12-components, it turns out that commutativity is not even valid for symmetric matrices.
(b) Both sides of the equation lead to the components

$$
\begin{aligned}
& {[(\underline{\underline{A B}}) \underline{\underline{C}}]_{i j}=\sum_{l}(\underline{\underline{A B}})_{i l} c_{l j}=\sum_{l}\left(\sum_{k} a_{i k} b_{k l}\right) c_{k j},} \\
& {[\underline{\underline{A}}(\underline{\underline{B C}})]_{i j}=\sum_{k} a_{i k}(\underline{\underline{B C}})_{k j}=\sum_{k} a_{i k}\left(\sum_{l} b_{k l} c_{l j}\right) .}
\end{aligned}
$$

Because of associativity of numbers, the parentheses can be set arbitrarily. Furthermore, the sum over $l$ can be positioned before $a_{i k}$ in the second expression because $a_{i k}$ does not depend on $l$. Hence, both expressions are equal to

$$
\sum_{k} \sum_{l} a_{i k} b_{k l} c_{l j}
$$

proving the statement for matrix multiplication. In effect, the associativity of the matrix products is reduced to the associativity of numbers (independence of which part of the product is calculated first). For products of three objects including vectors, the proof developed above is valid as well since this is just a special case of appropriately degenerated matrices.
(c) The relations $\underline{\underline{A}}(\vec{b}+\vec{c})=\underline{\underline{A}} \vec{b}+\underline{\underline{A}} \vec{c} \quad$ and $\quad \underline{\underline{A}}(\underline{\underline{B}}+\underline{\underline{C}})=\underline{\underline{A B}}+\underline{\underline{A C}}$ follow directly from the elementwise addition of vector and matrix components and from the distributivity of numbersfolgen. A direct calculation makes this explicit.
(d) $\left(\underline{\underline{A^{\prime}}}\right)^{\prime}=\underline{\underline{A}}$ since the transposition swaps rows and columns and a double swap reverts the original expression.
(e) We can show this by explicit summation of components. For the object "matrix times vector transposed", we have for each component $i i$ :

$$
\left(\vec{b}^{\prime} \underline{\underline{A^{\prime}}}\right)_{i}=\sum_{k} b_{k}\left(\underline{\underline{A^{\prime}}}\right)_{k i}=\sum_{k} b_{k} a_{i k}=\sum_{k} a_{i k} b_{k}=(\underline{\underline{A}} \vec{b})_{i}^{\prime}
$$

This means, the numerical values are identical. Furthermore, both products "transposed vector times matrix" and "(matrix times vector) transposed" result in a transposed vector. This means, also the type of mathematical object is identical.
For the object "matrix times matrix transposed", we obtain for each component $i j$ :

$$
\left(\underline{\underline{B}}^{\prime} \underline{\underline{A^{\prime}}}\right)_{i j}=\sum_{k}\left(\underline{\underline{B^{\prime}}}\right)_{i k}\left(\underline{\underline{A^{\prime}}}\right)_{k j}=\sum_{k} b_{k i} a_{j k}=\sum_{k} a_{j k} b_{k i}=(\underline{\underline{A B}})_{j i}=(\underline{\underline{A B}})_{i j}^{\prime} .
$$

(f) This identity can be shown by explicit calculation of the components. More elegantly, however, is applying the just derived transposition rule $(\underline{\underline{A B}})^{\prime}=\underline{\underline{B}}^{\prime} \underline{\underline{A}}^{\prime}$ together with the switching property of the transposition onto $\underline{\underline{A}}=\underline{\underline{X}}$ and $\underline{\underline{B}}=\underline{\underline{X^{\prime}}}$, respectively:

$$
\left(\underline{\underline{X}}^{\prime} \underline{\underline{X}}\right)^{\prime}=\underline{\underline{X}}^{\prime}\left(\underline{\underline{X^{\prime}}}\right)^{\prime}=\underline{\underline{X}}^{\prime} \underline{\underline{X}}
$$

(g) The unit matrix $\underline{\underline{E}}$ (sometimes denoted by $\underline{1}$ ) is a square matrix with ones on the diagonal, and zeroes, otherwise. It is the "neutral element" of the operation "matrix multiplication" and hence defines the inverse of square matrices. If this matrix is invertible or "regular" (notice that non-square matrices cannot be inverted in any case), following definition applies 1

$$
\underline{\underline{A}}^{-1} \underline{\underline{A}}=\underline{\underline{A A}}^{-1}=\underline{\underline{E}}
$$

Furthermore, we have, of course, $\underline{\underline{E^{\prime}}}=\underline{\underline{E}}$ which means

$$
\underline{\underline{E^{\prime}}}=\left(\underline{\underline{A}}^{-1} \underline{\underline{A}}\right)^{\prime}=\underline{\underline{A}}^{\prime}\left(\underline{\underline{A}}^{-1}\right)^{\prime}
$$

In the second line, we have used the transposition rule for matrix products derived earlier. Multiplying this equation from left ${ }^{2}$ by $\left(\underline{\underline{A^{\prime}}}\right)^{-1}$, we obtain for the lhs.

$$
\left(\underline{\underline{A^{\prime}}}\right)^{-1} \underline{\underline{E}}=\left(\underline{\underline{A^{\prime}}}\right)^{-1}
$$

[^0]The rhs. results in

$$
\left(\underline{\underline{A}}^{\prime}\right)^{-1}\left(\underline{\underline{A}}^{\prime}\left(\underline{\underline{A}}^{-1}\right)^{\prime}\right)=\left(\underline{\underline{A}}^{\prime}\right)^{-1} \underline{\underline{A}}^{\prime}\left(\underline{\underline{A}}^{-1}\right)^{\prime}=\left(\underline{\underline{A}}^{-1}\right)^{\prime}
$$

or $\left(\underline{\underline{A}}^{\prime}\right)^{-1}=\left(\underline{\underline{A}}^{-1}\right)^{\prime}$, q.e.d. Notice that, in the first step of evaluating the rhs., we made use of the associativity of matrix multiplication.

## Solution to Problem 2.2: Matrix Inversion

(a) The matrix multiplication cna be done efficiently "by hand" using Falk's scheme formalizing the multiplication rule

$$
(\underline{\underline{A B}})_{i j}=i \text {-th row of } \underline{\underline{A}} \text { times } j \text {-th column of } \underline{\underline{B}}
$$

For $2 \times 2$ matrices, this scheme reads

|  |  | e | f |
| :---: | :---: | :---: | :---: |
|  |  | g | h |
| a | b | $\mathrm{ae}+\mathrm{bg}$ | $\mathrm{af}+\mathrm{bh}$ |
| c | d | $\mathrm{ce}+\mathrm{cg}$ | $\mathrm{cf}+\mathrm{dh}$ |

. Here, the left factor (in the scheme at the upper left) equals $\underline{\underline{A}}^{-1}$, i.e.,

$$
e=\frac{d}{\operatorname{det} \underline{\underline{A}}}, \quad f=-\frac{b}{\operatorname{det} \underline{\underline{A}}}, \quad g=-\frac{c}{\operatorname{det} \underline{\underline{A}}}, \quad h=\frac{a}{\operatorname{det} \underline{\underline{A}}} .
$$

For example,

$$
\left(\underline{\underline{A A}}^{-1}\right)_{11}=a e+b g=\frac{a d-b c}{a d-c b}=1
$$

The other three elements are tretaed similarly with the result

$$
\underline{\underline{A A}}^{-1}=\left(\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right)=\underline{\underline{E}}
$$

q.e.d.
(b) Using Falk's scheme, we again prove the identity element by element:


For the 11 element of the product, we obtain

$$
\begin{equation*}
\frac{a(e i-f h)+d(c h-b i)+g(b f-c e)}{a e i+b f g+c d h-a f h-b d i-c e g}=1 \tag{1}
\end{equation*}
$$

The other elements are evaluated similarly.

## Solution to Problem 2.3: Vector and Matrix Derivatives

Derivative $\frac{\partial}{\partial \vec{\beta}}\left(\vec{\beta}^{\prime} \vec{a}\right)$ :

$$
\frac{\partial}{\partial \vec{\beta}}\left(\vec{\beta}^{\prime} \cdot \vec{a}\right) \stackrel{!}{=}\left(\begin{array}{c}
\frac{\partial}{\partial \beta_{0}}\left(\sum_{j} \beta_{j} a_{j}\right) \\
\vdots \\
\frac{\partial}{\partial \beta_{J}}\left(\sum_{j} \beta_{j} a_{j}\right)
\end{array}\right)=\left(\begin{array}{c}
a_{0} \\
\vdots \\
a_{J}
\end{array}\right)=\vec{a}
$$

Derivative $\frac{\partial}{\partial \vec{\beta}}\left(\vec{\beta}^{\prime} \underline{\underline{A}} \vec{\beta}\right)$ taking into account the product rule of differentiation:

$$
\begin{aligned}
\frac{\partial}{\partial \vec{\beta}}\left(\vec{\beta}^{\prime} \underline{\underline{A}} \vec{\beta}\right) & \stackrel{!}{=}\left(\begin{array}{c}
\frac{\partial}{\partial \beta_{0}}\left(\sum_{j} \sum_{k} \beta_{j} A_{j k} \beta_{k}\right) \\
\vdots \\
\frac{\partial}{\partial \beta_{J}}\left(\sum_{j} \sum_{k} \beta_{j} A_{j k} \beta_{k}\right)
\end{array}\right) \\
& =\left(\begin{array}{c}
\sum_{k} A_{0 k} \beta_{k}+\sum_{j} \beta_{j} A_{j 0} \\
\vdots \\
\sum_{k} A_{J k} \beta_{k}+\sum_{j} \beta_{j} A_{j J}
\end{array}\right) \\
& =\underline{\underline{A} \vec{\beta}}+\underline{\underline{A}}^{\prime} \vec{\beta}
\end{aligned}
$$


[^0]:    ${ }^{1}$ Notice that multiplication with the unit matrix is the single exception from the non-commutativity of matrix products.
    ${ }^{2}$ Because of the non-commutativity, one needs to specify if the matrix is multiplied "from left" or "from the right".

