

Sympy_DerivScalarMatrix

October 2, 2020

```
[1]: from sympy import diff, sin, exp, symbols, Function, Matrix, □  
     ↪ MatrixSymbol, FunctionMatrix, derive_by_array
```

```
[2]: from sympy import Symbol  
  
def var(letter: str, i: int, j: int) -> Symbol:  
    letter_ij = Symbol('{}_{}'.format(letter, i+1, j+1), □  
    ↪ is_commutative=True)  
    return letter_ij  
  
def func(i, j):  
    y_ij = Function('y_{}'.format(i+1,j+1))(*X)  
    return y_ij  
  
n,m,p = 3,3,2  
  
X = Matrix(n, m, lambda i, j: var('x', i, j)); X
```

```
[2]: 
$$\begin{bmatrix} x_{11} & x_{12} & x_{13} \\ x_{21} & x_{22} & x_{23} \\ x_{31} & x_{32} & x_{33} \end{bmatrix}$$

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```
[3]: #Y = MatrixSymbol(Function('y'), 2, 3); Matrix(Y)  
     #M = MatrixSymbol('M',2,2); Matrix(M)  
     #Y = Matrix(m, p, lambda i,j: Function('y_{}'.format(i+1,j+1))(X) ); Y  
  
Y = Matrix(m, p, lambda i,j: func(i, j)); Y
```

```
[3]:
```

$$\begin{bmatrix} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \mathbf{y}_{12}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \mathbf{y}_{21}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \mathbf{y}_{22}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \mathbf{y}_{31}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \mathbf{y}_{32}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \end{bmatrix}$$

0.0.1 Derivative of Scalar Function of a Matrix with Respect to the Matrix

Let $X = \{x_{ij}\}$ be a matrix of order $m \times n$ and let

$$y = f(X)$$

be a scalar function of X , so $y \in \mathbb{R}$ and $f: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$, Then we can define the derivative of y with respect to X as the following matrix of order $m \times n$:

$$\frac{\partial y}{\partial X} = \begin{pmatrix} \frac{\partial y}{\partial x_{11}} & \frac{\partial y}{\partial x_{12}} & \cdots & \frac{\partial y}{\partial x_{1n}} \\ \frac{\partial y}{\partial x_{21}} & \frac{\partial y}{\partial x_{22}} & \cdots & \frac{\partial y}{\partial x_{2n}} \\ \vdots & \vdots & & \vdots \\ \frac{\partial y}{\partial x_{m1}} & \frac{\partial y}{\partial x_{m2}} & \cdots & \frac{\partial y}{\partial x_{mn}} \end{pmatrix} = \left\{ \frac{\partial y}{\partial x_{ij}} \right\}$$

The matrix $\frac{\partial y}{\partial X}$ is called the gradient matrix.

[4]: `derive_by_array(Y[0,0], X)`

[4]:
$$\begin{bmatrix} \frac{\partial}{\partial x_{11}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{12}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{13}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \frac{\partial}{\partial x_{21}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{22}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{23}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \frac{\partial}{\partial x_{31}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{32}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{33}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \end{bmatrix}$$

0.0.2 Derivative of Matrix With Respect to Scalar Element of Matrix

Let $X = \{x_{ij}\}$ be a matrix of order $m \times n$ and let

$$y = f(X)$$

be a scalar function of X , so $y \in \mathbb{R}$ and $f: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$,

Also let the matrix $Y = \{y_{ij}(X)\}$ be of size $p \times q$.

Then we can define the derivative of Y with respect to an element

x in X as the following matrix of order $p \times q$:

$$\frac{\partial Y}{\partial x} = \begin{pmatrix} \frac{\partial Y}{\partial x} & \frac{\partial Y}{\partial x} & \cdots & \frac{\partial Y}{\partial x} \\ \frac{\partial Y}{\partial x} & \frac{\partial Y}{\partial x} & \cdots & \frac{\partial Y}{\partial x} \\ \vdots & \vdots & & \vdots \\ \frac{\partial Y}{\partial x} & \frac{\partial Y}{\partial x} & \cdots & \frac{\partial Y}{\partial x} \end{pmatrix} = \left\{ \frac{\partial y_{ij}}{\partial x} \right\}$$

[5]: `derive_by_array(Y, X[1-1,2-1])`

[5]:
$$\begin{bmatrix} \frac{\partial}{\partial x_{12}} \mathbf{y}_{11}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{12}} \mathbf{y}_{12}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \frac{\partial}{\partial x_{12}} \mathbf{y}_{21}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{12}} \mathbf{y}_{22}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \\ \frac{\partial}{\partial x_{12}} \mathbf{y}_{31}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) & \frac{\partial}{\partial x_{12}} \mathbf{y}_{32}(x_{11}, x_{12}, x_{13}, x_{21}, x_{22}, x_{23}, x_{31}, x_{32}, x_{33}) \end{bmatrix}$$