

Math for Deep Learning

What You Need to Know to Understand Neural Networks

by Ronald T. Kneusel

errata updated to print 3

Page	Error	Correction	Print corrected																		
5	... we can let NumPy choose the data type for us, or we can specify it explicitly we can let NumPy choose the data type for us, or we can specify it explicitly .	Pending																		
6	<table border="1"> <thead> <tr> <th>NumPy Name</th> <th>Equivalent C Type</th> <th>Range</th> </tr> </thead> <tbody> <tr> <td>float32</td> <td>float</td> <td>$\pm [1.175 \times 10^{-38}, 3.403 \times 10^{38}]$</td> </tr> <tr> <td>uint8</td> <td>unsigned char</td> <td>$[0, 255 = 2^8 - 1]$</td> </tr> </tbody> </table>	NumPy Name	Equivalent C Type	Range	float32	float	$\pm [1.175 \times 10^{-38}, 3.403 \times 10^{38}]$	uint8	unsigned char	$[0, 255 = 2^8 - 1]$	<table border="1"> <thead> <tr> <th>NumPy Name</th> <th>Equivalent C Type</th> <th>Range</th> </tr> </thead> <tbody> <tr> <td>float32</td> <td>float</td> <td>$\pm [1.175 \times 10^{-38}, 3.403 \times 10^{38}]$</td> </tr> <tr> <td>uint8</td> <td>unsigned char</td> <td>$[0, 255 = 2^8 - 1]$</td> </tr> </tbody> </table>	NumPy Name	Equivalent C Type	Range	float32	float	$\pm [1.175 \times 10^{-38}, 3.403 \times 10^{38}]$	uint8	unsigned char	$[0, 255 = 2^8 - 1]$	Print 2
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18	If there's no chance the something will happen, its probability is zero.	If there's no chance that something will happen, its probability is zero.	Pending																		
29	<pre>a = np.random.randint(0,364) b = np.random.randint(0,364)</pre> <p>The code simulates 100,000 random pairs of people, where the random integer in [0, 364] represents the person's birthday.</p>	<pre>a = np.random.randint(0,365) b = np.random.randint(0,365)</pre> <p>The code simulates 100,000 random pairs of people, where the random integer in [0, 365] represents the person's birthday.</p>	Pending																		
29	<pre>b = np.random.randint(0,364,m)</pre>	<pre>b = np.random.randint(0,365,m)</pre>	Pending																		
39	... which shows us correct way to compare conditional probabilities.	... which shows us the correct way to compare conditional probabilities.	Pending																		
39	<pre>s = np.random.randint(0,50,3)</pre>	<pre>s = np.random.choice(50,3,replace=False)</pre>	Pending																		
82	The top chart in Figure 4-4 shows the box plot for the three sets of exam scores in <code>!exams.npy!</code> .	The top chart in Figure 4-4 shows the box plot for the three sets of exam scores in <code>exams.npy</code> .	Pending																		

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119	Equation replacement	$\mathbf{a} \times \mathbf{b} = \ \mathbf{a}\ \ \mathbf{b}\ \sin(\theta) \hat{\mathbf{n}}$ $= (a_1 b_2 - a_2 b_1, a_2 b_0 - a_0 b_2, a_0 b_1 - a_1 b_0) \quad (5.6)$	Print 3
128	Equation replacement	$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \begin{bmatrix} 11 \\ 12 \\ 13 \end{bmatrix} = \begin{bmatrix} 74 \\ 182 \\ 290 \end{bmatrix}$	Print 3
131	for $n, m \in \mathbb{I}^+$ (positive integers) and where \mathcal{A} is a square matrix.	for $n, m \in \mathbb{Z}^+$ (positive integers) and where \mathcal{A} is a square matrix.	Pending
175	But $e^{\mathbf{x} \ln \mathbf{a}} = a^{\mathbf{x}}$, so we have . . .	But $e^{\mathbf{x} \ln \mathbf{a}} = a^{\mathbf{x}}$, so we have . . .	Print 3
183	For example, above, we saw that the partial derivative of $f(x, y) = \dots$	For example, above, we saw that the partial derivative of $f(x, y, \mathbf{t}, \mathbf{z}) = \dots$	Print 3
198	Equation replacement	$\frac{\partial \mathbf{F}}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial f_{00}}{\partial x} & \frac{\partial f_{01}}{\partial x} & \dots & \frac{\partial f_{0,m-1}}{\partial x} \\ \frac{\partial f_{10}}{\partial x} & \frac{\partial f_{11}}{\partial x} & \dots & \frac{\partial f_{1,m-1}}{\partial x} \\ \vdots & \vdots & \dots & \vdots \\ \frac{\partial f_{n-1,0}}{\partial x} & \frac{\partial f_{n-1,1}}{\partial x} & \dots & \frac{\partial f_{n-1,m-1}}{\partial x} \end{bmatrix}$	Print 3
201	Assume f accepts an m -element input and returns an n -element vector output.	Assume f accepts an m -element input and returns an n -element vector output.	Pending
236	Equation replacement	$f_0 : \begin{bmatrix} 4 & 11 & 8 \\ 9 & 8 & 1 \\ 15 & 0 & 6 \end{bmatrix} + \begin{bmatrix} 10 & 5 & 4 \\ 1 & -2 & -1 \\ -6 & -4 & -3 \end{bmatrix} = \begin{bmatrix} 14 & 16 & 12 \\ 10 & 6 & 0 \\ 9 & -4 & 3 \end{bmatrix} + 1 = \begin{bmatrix} 15 & 17 & 13 \\ 11 & 7 & 1 \\ 10 & -3 & 4 \end{bmatrix}$	Pending

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257	Equation replacement	$\frac{\partial E}{\partial \mathbf{x}} = \frac{\partial E}{\partial \mathbf{y}} \frac{\partial \mathbf{y}}{\partial \mathbf{x}}$ $= \left[\frac{\partial E}{\partial y_0} \frac{\partial y_0}{\partial x_0} \frac{\partial E}{\partial y_1} \frac{\partial y_1}{\partial x_1} \dots \right]^\top$ $= \left[\frac{\partial E}{\partial y_0} \sigma'(x_0) \frac{\partial E}{\partial y_1} \sigma'(x_1) \dots \right]^\top$ $= \frac{\partial E}{\partial \mathbf{y}} \odot \boldsymbol{\sigma}'(\mathbf{x}) \quad (10.10)$	Print 3
261	<pre>self.delta_w += np.dot(self.input.T, output_error)</pre>	<pre>self.delta_w += np.dot(weights_error)</pre>	Print 3
265	... so we reshape the training data from (60000, 196) to (60000,1,196) so we reshape the training data from (60000, 14,14) to (60000,1,196) ...	Pending
286	As before, we begin at $x = 0.75$...	We begin at $x = 0.75$...	Pending
300	Here, t , an integer starting at zero , is the timestep.	Here, t , an integer starting at one , is the timestep.	Pending
307	URL update	You can find them here: https://www.cs.toronto.edu/~binton/coursera_lectures.html	Print 2