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CHOO-DI Introduction - Trating Neural Networks

(model training)

TUTE ? ZENKOLVEZ VEGLASEN, CNNOL 18th Clarac of seal.

(LeNet)

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classes: lawweeter Update - Gradient Calculation

+ teamer gry very , other oder olfornen. forward & AN-> Dense-> Gigmord -> LOG-> CON-> Pooling -> CNN softmax

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RESIDENTIFE PONDWEGET IT CHE f(x; (9)) + oum got country of order the began ever. A DA SHARE THANH THE Pavameter = 45 75 25 chier. O On the steps dem. unst- netn est get. (A) Y CONTROL 2545 2015014. 141E1, acon. Tengor Itle. . Zeroth Order tensor (obten) Scolar O. 6 CR + Cartegian Product IR atto: RXR-) P: C5(10) E/RX112 a.b: RXR+ 1R B= 50 (3 BXB={(0,0),(0,1),((,0),(1,1)}

• Fürst-Order tensor (Nextor) (200th order operation = 200)

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am: RXRMXn > RMXn

M+N: Rmxn X Ruxn + Ruxn MON: RMEN XRAKEN RMXN M. W: RMEN & R"=R"

MO: RMXN X RNXO + RMXO

Second-order Tengor (Matrix)

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ME (2 2 4 4)

MINGRIMAN DE RINXO

AM: RXRIMXN > RINXO

· Thurd-Order Tennor Quartions. (MNIGI) RAD OER MINERMENTO WIN: BMANKO X BMKNKO - BMKNKO MON: RIMKNED RIMKNED RIMKNED - Dota Get (X Data) THE (() () () () ! THEREZ HOW VECTORE THE TORM & (20th STEPLANT STENT SEE) 3/4 1 (XC)) = (x(1)x2 ... x(2) leagen of caput $\rightarrow li$ 8/492 (() = (() () () () () () = (\(\langle \(\langle \) \

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CHOLO2 Artificial Neuman - Affine Function with one Feature. ZER Weighted Sum Z=KW) Z= (X-W+b. . After transformation 2=1W+b Affine Eunotions With n Features Weglited Sum Z= William + Wwxn = () TX = () TX W: 25952 $\frac{1}{12} = \begin{pmatrix} \frac{1}{12} & \frac{1}{12} \\ \frac{1}{12$ After transformation Z= CATW+b

Activation Function Sigmoid $g(x) = \sigma(x) = \frac{1}{1+e^{-x}}$ $\tau(x) = \frac{1}{1+e^{-x}}$ g (x)= ReLU(x)= max (O,x) Artitical News ずら) る(か)なけ 2(マラいり) (x) CRIXL: = g(tx)tutb) Wentiboach in Artificial Neurons よ((水が,かり) (x) butch Gize; will batch

Managardh in AN Q(1) (2) (2) (2) (2) > f((&)(3-12.6) a = 9 ((1xca)). 2 +6 microbatch; wibz obst 0(N) 9(X(N) = 3+6) 092KMESO Setch Hitx 女心之神的鬼. (minipatchof society with the Harlismotel) f(a;0)=1x+0 十0 三次的 编(包括数) f(2;0)=2+19 +(10; 0)= 10HO f(x1;2,6) orther when got MALVIYZ Usol ->mirrobatch vivilately weight / Bias > Affiretunction - Activation Function XTE RNXLi Z=Xwth ARR NXI ZERNXI

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(9(3))

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