

PokeNet: Predicting Pokemon Card Features Through Deep Learning

Eddy Albaran and Piper Keyes {albaran, pckeyes}@stanford.edu
CS230 Spring 2018



Introduction

We aim to develop Deep Learning networks (PokeNet, collectively) to predict various features of Pokemon cards. Specifically, our models predict type, hit points (HP), and card price. We employed three architectures (5 layer neural net, convolutional neural net, and DenseNet) to accomplish these tasks. Ultimately, we hope to expand the functionality of PokeNet to create a real life version of the famed and beloved Pokedex, fulfilling a life-long dream of many millennials that aspired to become Pokemon Masters.

Predicting Pokemon Type

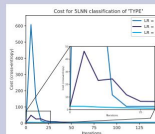
5-layer Neural Net

Parameter tuning

Learning rate	Iteration 1: identify optimal layers				Iteration 2: address bias by increasing epoch number				Iteration 3: address variance with L2 norm			
	5.0e-3	5.0e-4	5.0e-5	5.0e-6	5.0e-3	5.0e-4	5.0e-5	5.0e-6	5.0e-3	5.0e-4	5.0e-5	5.0e-6
Epoch num	150	150	250	350	150	250	350	350	150	250	350	350
Beta value					0.001	0.001	0.001	0.001	0.001	0.001	0.001	0.004
Train accuracy	0.146	0.930	0.959	0.963	0.979	0.984	0.996	0.997	0.997	0.999	0.999	0.995
Dev accuracy	0.158	0.914	0.953	0.951	0.951	0.959	0.971	0.978	0.976	0.971	0.973	0.973
Final cost	2.334	0.370	0.225	0.178	0.104	0.101	0.050	0.018	0.114	0.780	0.134	

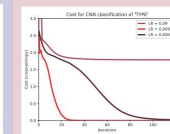
Final model (5e-4 lr, 350 epochs, 0.001 beta) test accuracy: 0.9755

Network training



Convolutional Neural Net

Network training



Train: 0.62
Dev: 0.54
Train: 0.97
Dev: 0.97
Train: 0.95
Dev: 0.94

Saliency mapping



P_{Water} = 0.94 P_{Fire} = 0.79 P_{Grass} = 0.69
P_{Electric} = 0.01

Dataset and Preprocessing

We scraped our dataset from pkmncards.com and rescaled the card image files to 224x224 pixels. We vectorized the images and concatenated the vectors into a single X matrix. Y labels were also vectorized. We collected all 9791 Pokemon cards in existence and divided them into Train/Dev/Test sets as follows: Train = 7832 cards (~80%), Dev = 979 cards (~10%), Test = 980 cards (~10%).

Earlier sets examples



Later sets examples



Predicting Pokemon HP

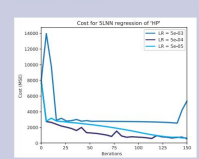
5-layer Neural Net

Parameter tuning

Learning rate	5.0e-3	5.0e-4	5.0e-5	5.0e-6
	Epoch num	150	150	150
Train mse	61.598	23.804	24.732	18.730
Dev mse	61.097	23.688	24.755	20.146
Final cost	5367.1	577.6	661.8	363.9

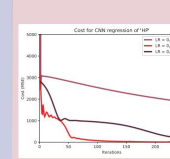
Final model (0.0005 lr, 250 epochs) test rmse: 20.819

Network training



Convolutional Neural Net

Network training



Train: 43.23
Dev: 42.69
Train: 8.57
Dev: 8.59
Train: 13.39
Dev: 15.99

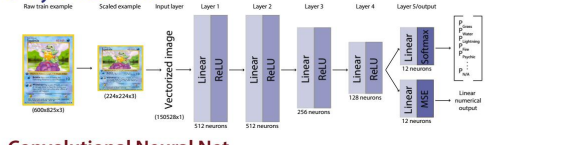
Saliency mapping



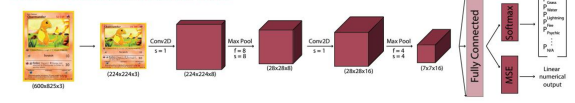
Pred HP: 93.2 True HP: 100
Pred HP: 47.1 True HP: 30

Architectures

5-layer Neural Net



Convolutional Neural Net



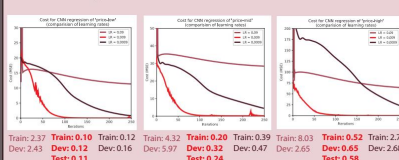
DenseNet-121



Predicting Card Price

Convolutional Neural Net

Network training



Train: 2.37 Train: 0.10 Train: 0.12 Train: 4.32 Train: 0.30 Train: 0.39 Train: 0.03 Train: 0.52 Train: 2.79
Dev: 2.43 Dev: 0.12 Dev: 0.16 Dev: 5.97 Dev: 0.32 Dev: 0.47 Dev: 2.65 Dev: 0.65 Dev: 2.68

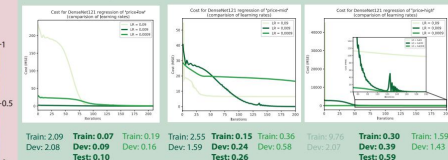
Saliency mapping



Pred mid price: \$49.90
True mid price: \$64.90

DenseNet-121

Network training



Train: 2.09 Train: 0.07 Train: 0.19 Train: 2.55 Train: 0.15 Train: 0.36 Train: 0.39 Train: 1.59
Dev: 2.08 Dev: 0.09 Dev: 0.16 Dev: 1.59 Dev: 0.24 Dev: 0.58 Dev: 3.03 Dev: 1.43
Test: 0.10 Test: 0.10 Test: 0.26 Test: 0.26 Test: 0.59 Test: 0.59

Pred mid price: \$51.27

References

- [1] Huang, Guo, Liu, Zhuang, van der Maaten, Laurens, Weinberger, Kilian Q. DenseNet: Connected Convolutional Networks, 2018.
- [2] <https://github.com/flyyufelix/DenseNet-Keras>
- [3] Zhou, Bolei, Khosla, Aditya, Lapedriza, Agata, Oliva, Aude; Torralba, Antonio. Learning Deep Features for Discriminative Localization, 2016.
- [4] <http://pkmncards.com>
- [5] https://github.com/philipperemy/tensorflow-class-activation-mapping/blob/master/class_activation_map.py
- [6] CS230 programming assignments "TensorFlow Tutorial" and "Convolutional model Application v1"