# Digit Recognizer

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## Purpose

The purpose of this machine learning algorithm is to be able to determine what number is being depicted from a picture of a handwritten digit.

# Data Layout

• Each picture is divided into 784 pixels. Each pixel is given a value from 0-255 to depict how full each pixel is. If a pixel is completely black, it will be given the value of 255 and if it's completely white (empty), it will be given the value of 0.

1	XZ	YA	YB	YC	YD	YE	YF	YG	YH	YI	YJ	YK	YL	YM	YN	YO	YP	YQ	YR
1 p	ixel648	pixel649	pixel650	pixel651	pixel652	pixel653	pixel654	pixel655	pixel656	pixel657	pixel658	pixel659	pixel660	pixel661	pixel662	pixel663	pixel664	pixel665	pixel666
2	0	0	0	0	214	218	95	0	0	0	0	0	0	0	0	(	(	) (	0
3	0	0	0	0	0	8	76	146	254	255	254	255	146	19	15	C	(	) (	0 0
4	0	0	0	0	0	0	0	0	63	254	254	62	0	0	0	C	(	) (	0 0
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	(	186	159	0
6	0	0	0	1	18	129	208	253	253	253	253	159	129	90	4		(	) (	0 0
7	0	0	0	0	0	23	64	158	200	174	61	0	0	0	0	C	(	) (	0 0
8	0	0	0	0	0	0	0	0	0	152	253	82	0	0	0	C	(	) (	0 0
9	0	0	94	248	209	73	12	0	0	0	0	0	0	42	147	252	136	5 9	0
10	0	0	0	0	79	243	234	254	253	253	216	117	0	0	0	C	(	) (	0 0
11	0	0	67	50	176	148	78	16	0	12	12	0	0	0	0	C	(	) (	0 0
12	0	7	186	252	227	184	191	252	252	252	252	253	240	50	0	(	(	) (	0
13	0	0	0	0	33	218	252	252	192	141	14	0	0	0	0	C	(	) (	0 0
14	0	0	0	0	0	0	0	0	9	254	87	0	0	0	0	(	(	) (	0
15	0	0	0	0	0	0	0	255	255	255	255	255	255	255	255	191	. (	) (	0 0
16	0	100	221	252	252	253	127	112	112	112	0	0	0	0	0	(	(	) (	0
17	0	0	0	0	0	0	0	45	253	66	0	0	0	0	0	C	(	) (	0 0
18	0	0	0	0	0	33	179	241	253	253	253	253	253	253	246	179	44	1 (	0
19	0	0	45	233	253	253	232	120	114	0	0	0	0	0	0	(	(	) (	0 0
20	0	0	0	0	0	0	1	191	226	27	0	0	0	0	0	C	(	) (	0
21	0	0	0	0	0	0	0	0	25	91	165	252	252	63	7	28		3 (	0 0
22	0	0	0	207	169	83	174	242	230	80	4	0	0	0	0	C	(	) (	0 0
23	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	C	(	) (	0 0
24	0	0	0	0	62	252	210	128	252	62	0	0	0	0	0	C	(	) (	0 0
25	0	0	0	0	0	11	23	22	107	137	137	23	22	22	14	C	(	) (	0 0
26	0	0	0	0	0	0	94	238	238	238	240	112	106	17	0	(	(	) (	0 0
27	145	248	253	212	229	253	214	199	106	42	0	0	0	0	0	C	(	) (	0 0
28	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	(	(	) (	0 0

# First Hypothesis and Problems

- Use a Multi-layer Perceptron with one hidden layer to iterate through the data using an eta between .1 and 1 and neurons between 1 and 150 which were compared through 10-fold cross-validation.
- The training set supplied by Kaggle was too large so I shank it to 1000 entries
- The algorithm took approximately 9 hours to complete

# **Conclusions from First Experiment**

- The best solution the Multi-Layer Perceptron found was with 122 neurons and an eta of 0.1.
- The cross-validation score came to about -6357.587 and the CV E<sub>out</sub> came to about .28486
- Although the CV E<sub>out</sub> is better than randomly guessing, it is still a bad method of guessing the digits
- Poor computation time and poor results lead me to use another algorithm

## Second Hypothesis and Problems

- The next algorithm I tried to implement was a K-Nearest-Neighbors algorithm that just used all of the default parameters (n\_neighbors = 5).
- The computation time using the whole dataset was a little more than an hour
- To test I shrunk the dataset to only 5000 entries

# **Conclusions from Second Experiment**

- The solution using KNN was much better than the solution given by the Multi-Layer Perceptron, giving me a CV E<sub>out</sub> of .928305
- This is a great score, however computation time for the entire dataset is a little long

#### **Overall Conclusions and Difficulties**

- The KNN neighbors approach yielded a much better solution than the Multi-Layer Perceptron
- One possible explanation for this could have incorrect implementation of the Multi-Layer Perceptron
- Possibly looking into other algorithms to compare against the KNN algorithm is a way to move forward to try and maximize the CV E<sub>out</sub> and improve computation time