





Introduction

- The goal was to create a machine learning algorithm to detect interest points on pictures of shoe treads, starting with corners
- Current corner detection methods are only good at picking up on sharp, well defined corners which shoeprints rarely have
- In the future, this type of technique could be used to aid further analysis by mapping the corner points onto shoe treads before more advanced techniques could work

Methods

- We first manually marked shoeprint photos with dots on the corners, and sampled from these images to create our dataset of corner pixels and non-corner pixels
- Sampling involved taking pixels labeled corners, and the surrounding areas, and non corner pixels with surrounding area
- We then experimented with different neural network structures to optimize accuracy on testing data that the network hasn't seen
- Examples of hyperparameters we checked are learning rates, number of training cycles, image size, and layer numbers
- The network was constructed with Python code, and the PyTorch package





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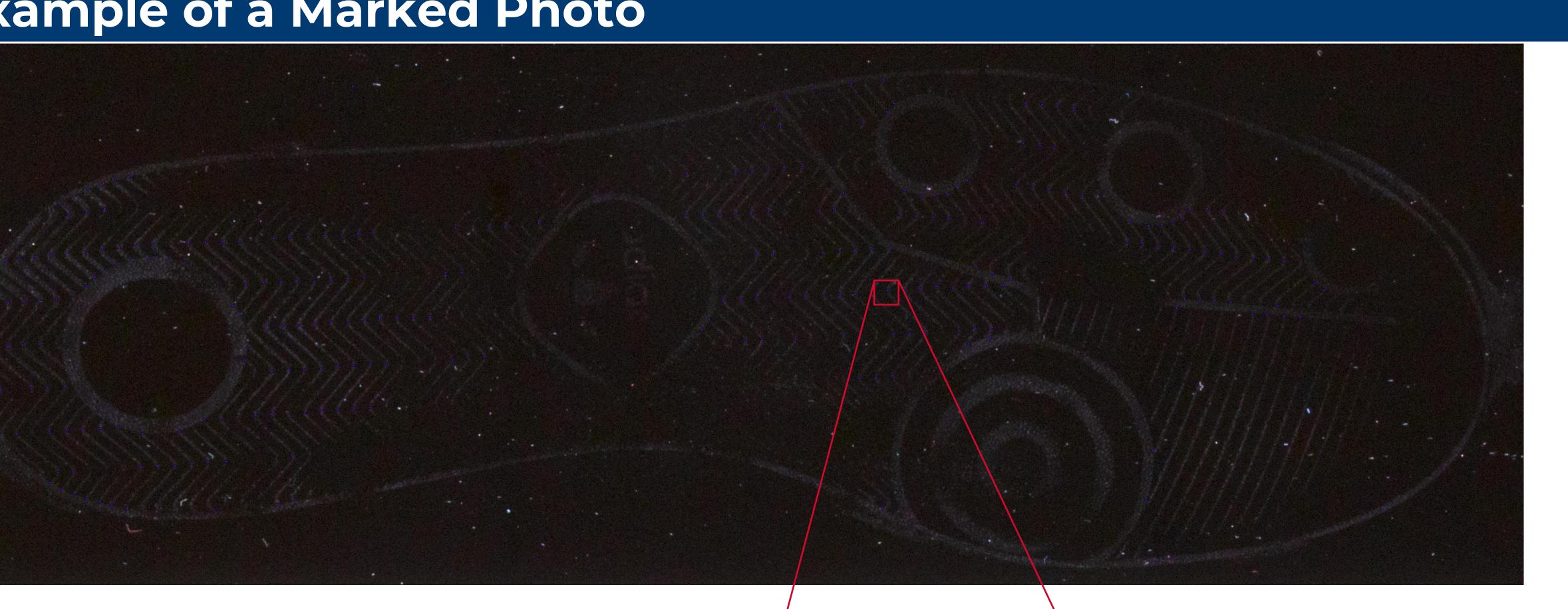
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Using Neural Networks to Identify Interest Points in Shoe Treads

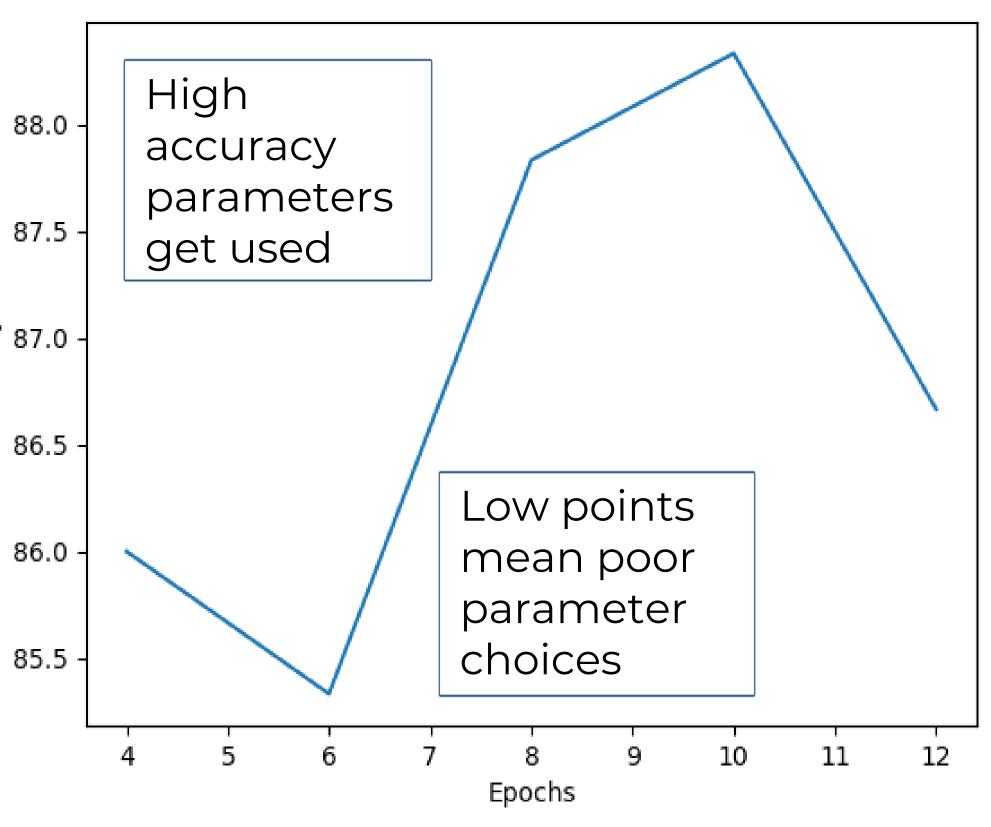
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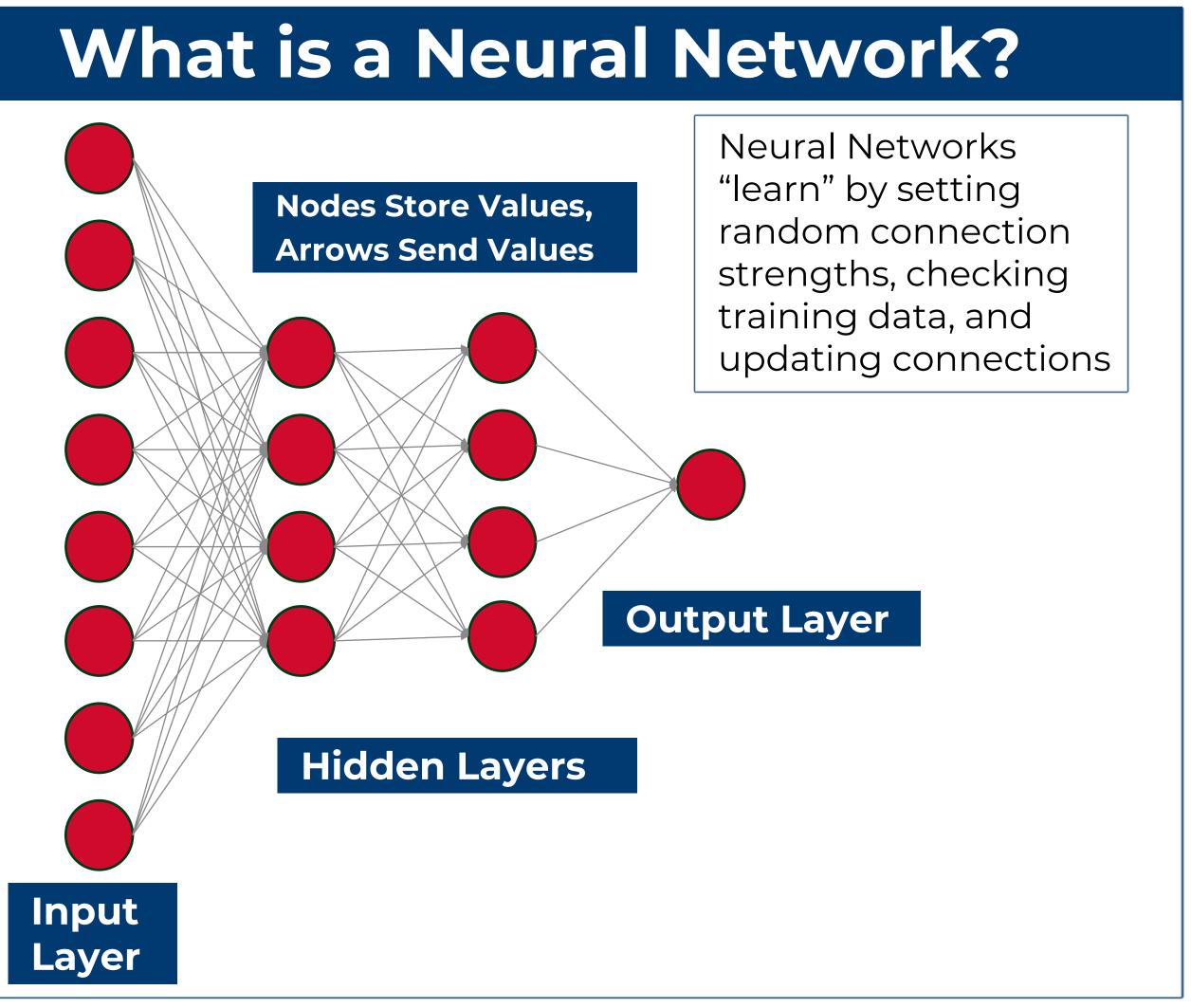
Example of a Marked Photo



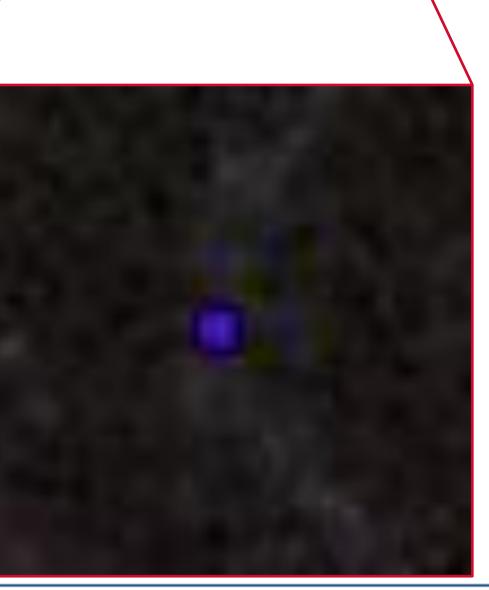
This square is a piece of data given to the network. The network must detect the fuzzy white lines that create a corner.

Parameter Data





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Example Code

				ᇢ main 🗸	æ	:
👌 my_net	twork.py ×	🗬 create_dataset.py	🍦 load_dataset.py	🍦 sample_network.py		
14	class Net(n	nn.Module):				
15	defi	<pre>init(self, in_shape):</pre>				
16	sup	per()init()				
17						
18	sel	f.features = nn.Sequer	ntial(
19			l, out_channels=3, k	ernel_size=5, padding=2),		
20		nn.BatchNorm2d(3),				
21		nn.MaxPool2d(2, 2),				
22		nn.ReLU(inplace=True)				
23						
24						
25	sel	f.classification = nn.	Sequential(
26		nn.Linear(108, 66),				
27		<pre>nn.ReLU(inplace=True) nn.kincen(((</pre>	1			
28		nn.Linear(66, 66),				
29 30		<pre>nn.ReLU(inplace=True) nn Linean(66 1)</pre>	1			
30 31		nn.Linear(66, 1), nn.Sigmoid()				
32						
33	,					
34	def for	<pre>rward(self, x):</pre>				
35		<pre>self.features(x)</pre>				
36		<pre>x.reshape(x.shape[0],</pre>	-1)			
37		<pre>self.classification(x</pre>				
38		turn x				

Results

- Largely successful, creating a model with 95% accuracy on data it had never seen before
- Simple models worked best overall, allowing for strong outcomes while preventing overfitting and speeding up computation times
- A single convolutional layer followed by 2 fully connected linear layers worked best

Conclusion

This model will need to be continuously improved, because shoeprint pictures contain thousands of pixels and marking even just 5% of them incorrectly means hundreds of wrong marks. What we have now will work as a powerful starting point for this future research.

After interest points can be quickly marked by a network, more advanced statistical comparisons can be made.