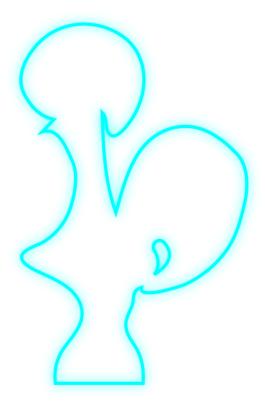
11/2/2020 Tiny Darknet

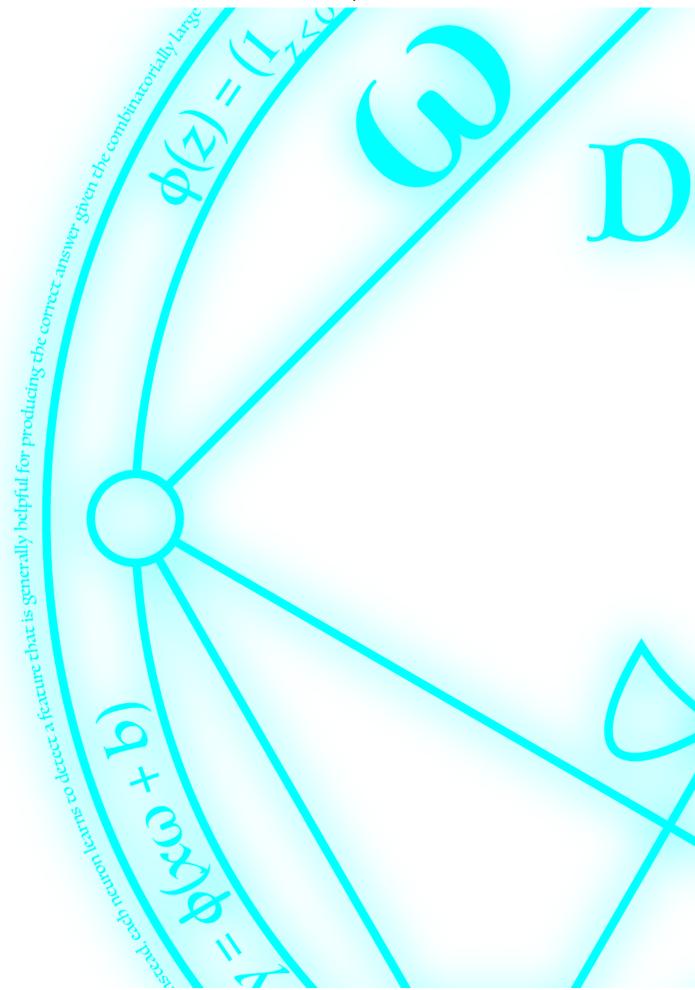
• home



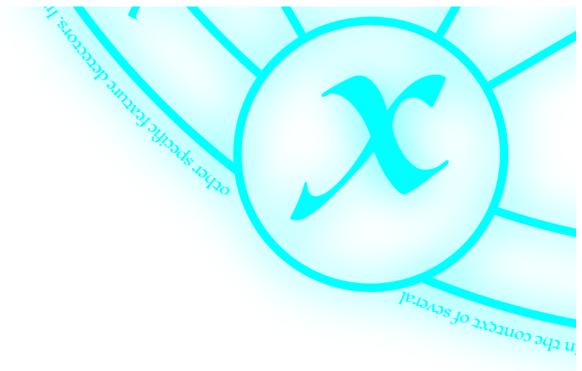


coq tactics

publications projects résumé



11/2/2020 Tiny Darknet



Tiny Darknet

I've heard a lot of people talking about SqueezeNet.

SqueezeNet is cool but it's JUST optimizing for parameter count. When most high quality images are 10MB or more why do we care if our models are 5 MB or 50 MB? If you want a small model that's actually FAST, why not check out the <u>Darknet reference network</u>? It's only 28 MB but more importantly, it's only 800 million floating point operations. The original <u>Alexnet</u> is 2.3 billion. Darknet is 2.9 times faster and it's small and it's 4% more accurate.

So what about SqueezeNet? Sure the weights are only 4.8 MB but a forward pass is still 2.2 billion operations. Alexnet was a great first pass at classification but we shouldn't be stuck back in the days when networks this bad are also this slow!

But anyway, people are super into SqueezeNet so if you really insist on small networks, use this:

Tiny Darknet

Model	Top-1	Top-5	Ops	Size
AlexNet	57.0	80.3	2.27 Bn	238 MB
Darknet Reference	61.1	83.0	0.81 Bn	28 MB
SqueezeNet	57.5	80.3	2.17 Bn	4.8 MB
Tiny Darknet	58.7	81.7	0.98 Bn	4.0 MB

The real winner here is clearly the Darknet reference model but if you insist on wanting a small model, use Tiny Darknet. Or train your own, it should be easy!

Here's how to use it in Darknet (and also how to install Darknet):

git clone https://github.com/pjreddie/darknet
cd darknet
make
wget https://pjreddie.com/media/files/tiny.weights
./darknet classify cfg/tiny.cfg tiny.weights data/dog.jpg

Hopefully you see something like this:

data/dog.jpg: Predicted in 0.160994 seconds. malamute: 0.167168 Eskimo dog: 0.065828 dogsled: 0.063020 standard schnauzer: 0.051153

Siberian husky: 0.037506

Here's the config file: $\underline{\text{tiny.cfg}}$

The model is just some 3x3 and 1x1 convolutional layers:

11/2/2020 Tiny Darknet

<pre>layer filters size</pre>	> 112 x 112 x 16 > 112 x 112 x 32 > 56 x 56 x 32 > 56 x 56 x 16
•	> 112 x 112 x 32 > 56 x 56 x 32 > 56 x 56 x 16
2 conv 32 3 x 3 / 1 112 x 112 x 16 -	> 56 x 56 x 32 > 56 x 56 x 16
	> 56 x 56 x 16
3 max 2 x 2 / 2 112 x 112 x 32 -	
4 conv 16 1 x 1 / 1 56 x 56 x 32 -	
5 conv 128 3 x 3 / 1 56 x 56 x 16 -	> 56 x 56 x 128
6 conv 16 1 x 1 / 1 56 x 56 x 128 -	> 56 x 56 x 16
7 conv 128 3 x 3 / 1 56 x 56 x 16 -	> 56 x 56 x 128
8 max 2 x 2 / 2 56 x 56 x 128 -	> 28 x 28 x 128
9 conv 32 1 x 1 / 1 28 x 28 x 128 -	> 28 x 28 x 32
10 conv 256 3 x 3 / 1 28 x 28 x 32 -	> 28 x 28 x 256
11 conv 32 1 x 1 / 1 28 x 28 x 256 -	> 28 x 28 x 32
12 conv 256 3 x 3 / 1 28 x 28 x 32 -	> 28 x 28 x 256
13 max 2 x 2 / 2 28 x 28 x 256 -	> 14 x 14 x 256
14 conv 64 1 x 1 / 1 14 x 14 x 256 -	> 14 x 14 x 64
15 conv 512 3 x 3 / 1 14 x 14 x 64 -	> 14 x 14 x 512
16 conv 64 1 x 1 / 1 14 x 14 x 512 -	> 14 x 14 x 64
17 conv 512 3 x 3 / 1 14 x 14 x 64 -	> 14 x 14 x 512
18 conv 128 1 x 1 / 1 14 x 14 x 512 -	> 14 x 14 x 128
19 conv 1000 1 x 1 / 1 14 x 14 x 128 -	> 14 x 14 x1000
20 avg 14 x 14 x1000 -	> 1000
21 softmax	1000
22 cost	1000