



What is Knowledge Distillation?



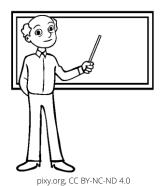
- knowledge distillation is a method for model compression
- first introduced by Hinton et al. in 2015 [1]

What is Knowledge Distillation?



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- knowledge of a teacher model is transferred/distilled into a student model

Teacher



Student



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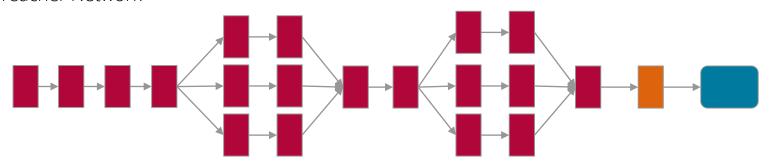
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Knowledge Distillation in Neural Networks





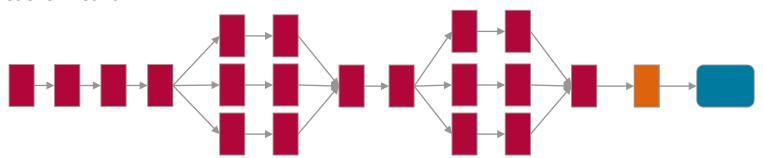


Convolution
Fully Connected
Softmax

Knowledge Distillation in Neural Networks



Teacher Network



trained by `hard` labels and softmax cross entropy



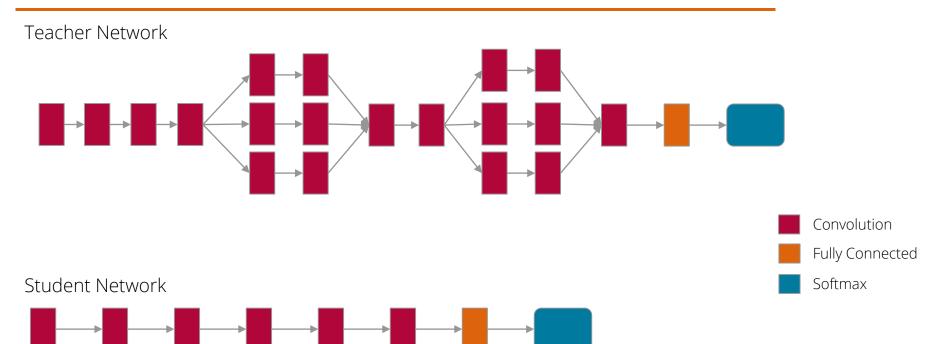
			1 3	
*	dog	0	[1, 0, 0, 0]	
	cat	1	[0, 1, 0, 0]	
	car	2	[0, 0, 1, 0]	
	ship	3	[0, 0, 0, 1]	

Convolution
Fully Connected

Softmax

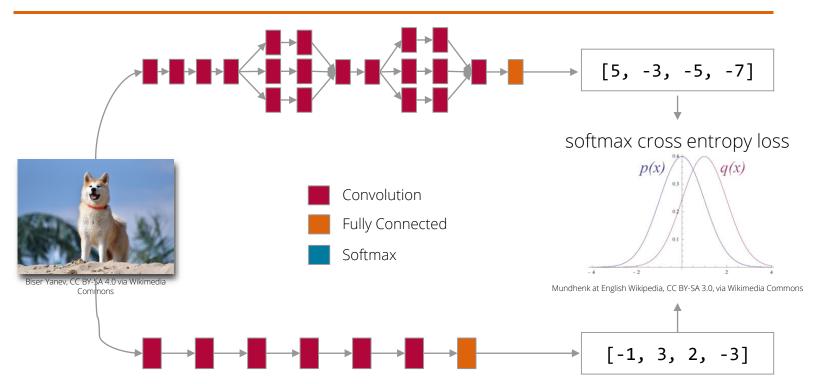
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Training the Student





Case Study ImageNet Training



Method	Parameters	Model Size	Top 1 Accuracy
Teacher (ResNet-152)	60,344,232	244 MB	77.98%

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Case Study ImageNet Training



Method	Parameters	Model Size	Top 1 Accuracy
Teacher (ResNet-152)	60,344,232	244 MB	77.98%
Student (ResNet-50)	25,610,216	104 MB	76.32%
Knowledge Distillation	25,610,216	104 MB	77.75%

Why does it work?



- models are often overparameterized
- soft targets contain more information about data than hard labels
 - i.e. a dog is more related to a cat, than a dog to a car
- soft targets also have less gradient variance
 - → smoother and easier training

Conclusion



- model distillation is a method for model compression
 - utilizes fact that large models are often overparameterized
- model distillation involves teacher and student
 - student learns based on soft labels produced by the teacher
- model distillation reduces energy usage of AI
- case study:
 - model compression of more than factor 2 possible
 - accuracy loss is minimal (0.23%)
- can also be used for compression of model ensembles

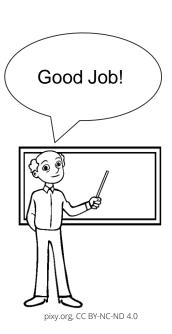


Chart 13