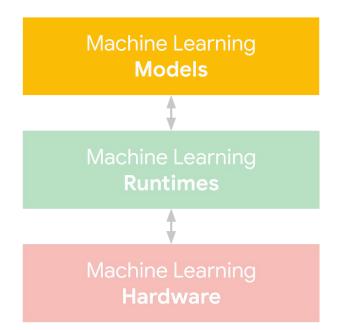
What are the Challenges for TinyML?

Part D





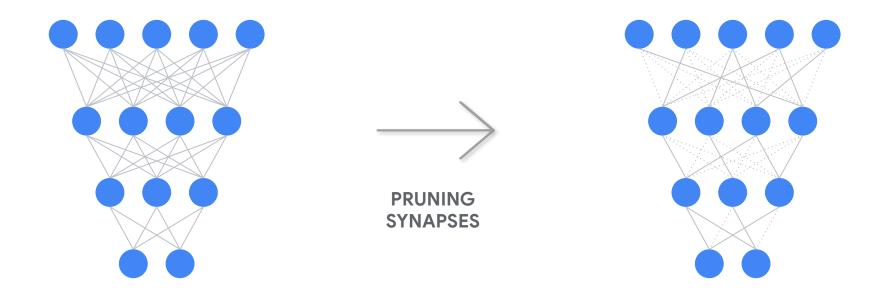
Model Compression Techniques

Pruning

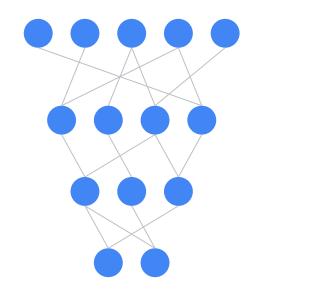
Quantization Knowledge Distillation

•••

Pruning

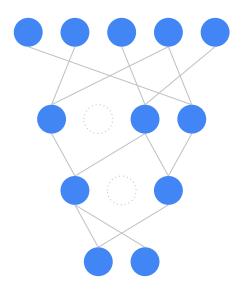


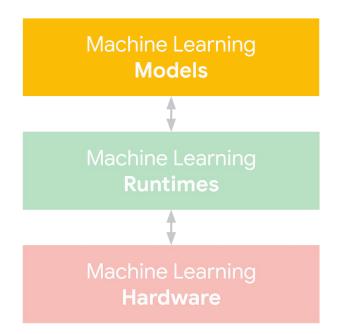
Pruning





PRUNING NEURONS





Model Compression Techniques

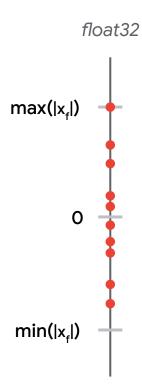
Pruning

Quantization

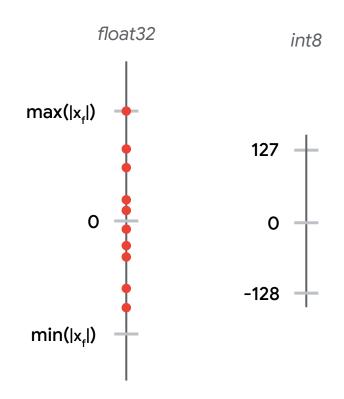
Knowledge Distillation

•••

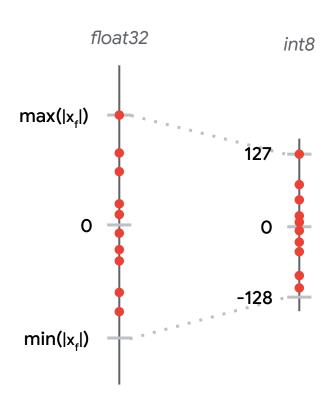
Quantization

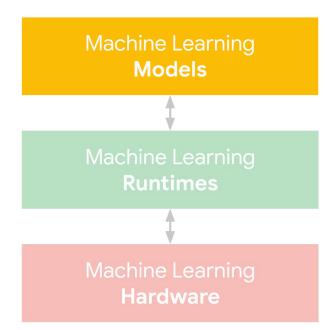


Quantization



Quantization





Model Compression Techniques

Pruning

Quantization

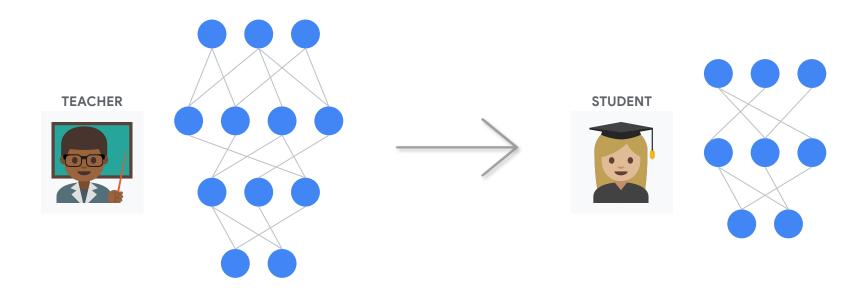
Knowledge Distillation

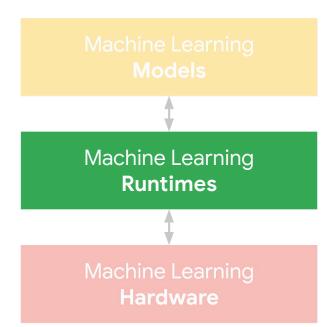
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Knowledge Distillation



Knowledge Distillation













Less memory

Less compute power

Only focused on inference



Key Differences



TensorFlow Lite

Topology

Weights

Binary Size

Distributed Compute

Developer Background

Variable

Variable

Unimportant

Needed

ML Researcher **Fixed**

Fixed

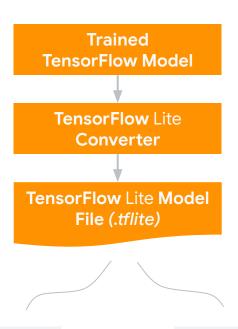
High Priority

Not Needed

Application Developer



Architecture









Even less memory

Even less compute power

Also, only focused on inference





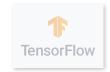
Train a model

Convert model

Optimize model at Edge

Deploy model at Edge

Make inferences at Edge





Convert model

Optimize model

Deploy model at Edge Make inferences at Edge





Convert model

Optimize model

Deploy model at Edge Make inferences at Edge





Convert model

Optimize model

Deploy model at Edge Make inferences at Edge







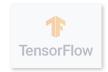








Microcontroller





Convert model

Optimize model

Deploy model at Edge Make inferences at Edge



