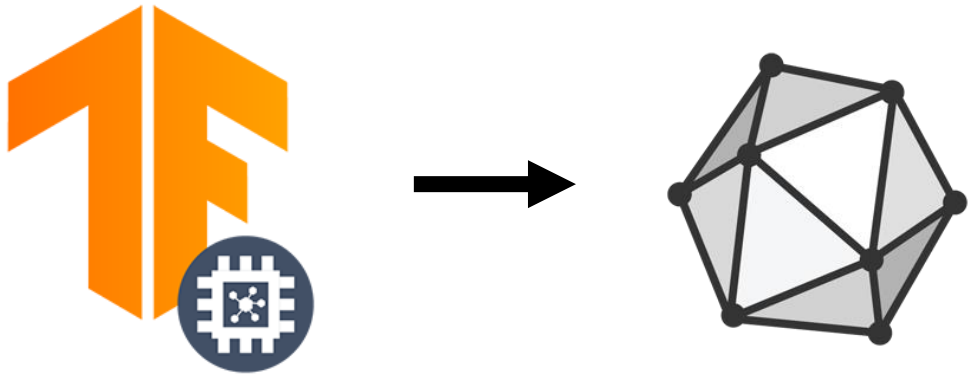
TensorFlow Lite to ONNX Conversion

ONNX Runtime Mobile

TOM WILDENHAIN

SCOTT MCKAY

MICROSOFT



TensorFlow Lite to ONNX

TOM WILDENHAIN SOFTWARE ENGINEER, ONNX CONVERTERS TEAM

MICROSOFT

TensorFlow

- Used for inference and training
- > 1,000 ops
- Already have conversions for many ops

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TFLite

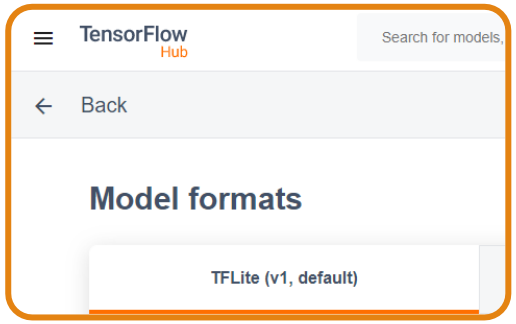
- Lightweight runtime used for inference
- ~130 ops
- Models created from TensorFlow

ABS
ADD_N
ARG_MAX
ARG_MIN
AVERAGE_POOL_2D
BATCH_TO_SPACE_ND
CAST
CEIL
CONCATENATION
CONV_2D
COS
CUMSUM
DEPTH_TO_SPACE
DEPTHWISE_CONV_2D
DEQUANTIZE
ELU
EQUAL
EXP
EXPAND_DIMS
FILL
FLOOR
FLOOR_DIV
FLOOR_MOD
GATHER
NEG
NON_MAX_SUPPRESSION_V4
NON_MAX_SUPPRESSION_V5
NOT_EQUAL
ONE_HOT
PACK
PAD
PADV2
POW
QUANTIZE
RANGE
RANK
REDUCE_ANY
REDUCE_MAX
REDUCE_PROD
RELU
RELU6
RESHAPE
RESIZE_BILINEAR
RESIZE_NEAREST_NEIGHBOR
REVERSE_SEQUENCE
REVERSE_V2
ROUND
RSORT
UNPACK
WHERE
WHILE
ZEROS_LIKE
FULLY_CONNECTED
ADD
DIV
MUL
SUB
BATCH_MATMUL
BIDIRECTIONAL_SEQUENCE_LSTM
BIDIRECTIONAL_SEQUENCE_RNN
BROADCAST_TO
CALL
CALL_ONCE
CONCAT_EMBEDDINGS
CUSTOM
DELEGATE
DENSIFY
EMBEDDING_LOOKUP
EMBEDDING_LOOKUP_SPARSE
FAKE_QUANT
HARD_SWISH
HASHTABLE_LOOKUP

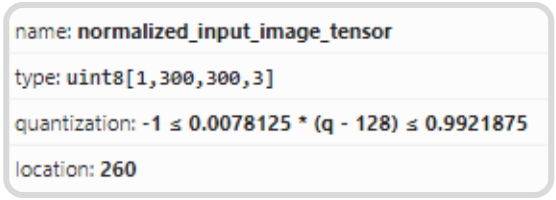
...

TFLite Conversion

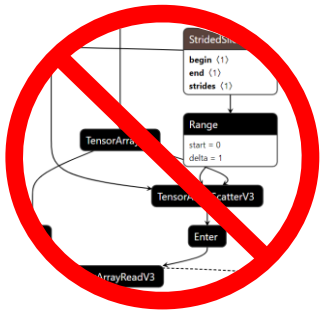
```
pip install tf2onnx
python -m tf2onnx.convert --tflite ssdmobilenet.tflite --output ssdmobilenet.onnx --opset 13
```



Some models are only available for TFLite



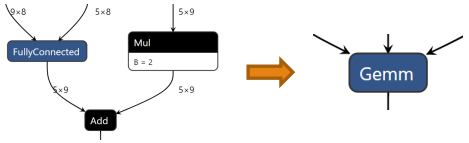
Automatic quantization support!



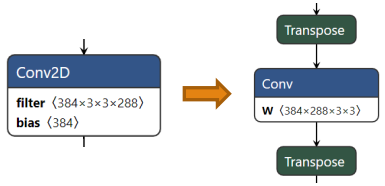
TFLite models are often cleaner

Conversion Process

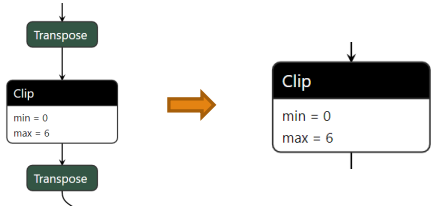
1. Rewriters
   - Convert op patterns



2. Handlers
   - Convert individual ops



3. Optimizers
   - Remove unnecessary ops



Conversion Process

1. **Rewriters**
   - **Convert op patterns**
2. Handlers
   - Convert individual ops
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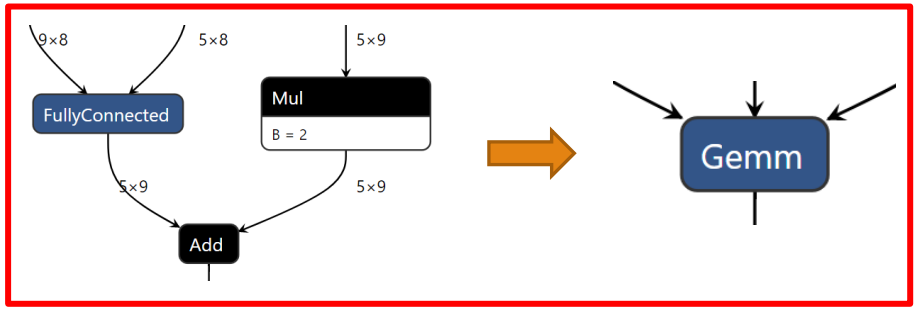

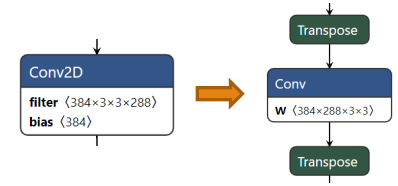

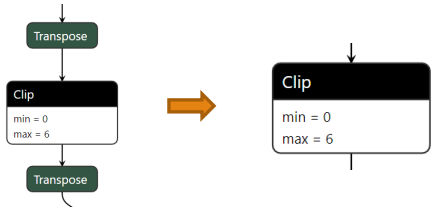

Conversion Process

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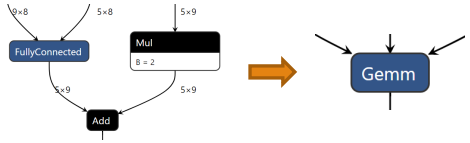

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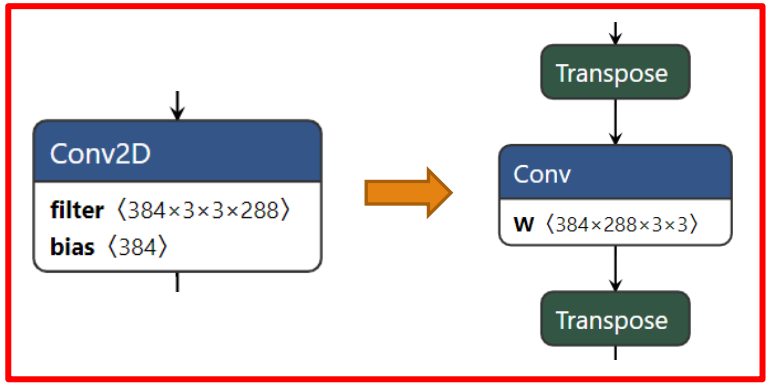

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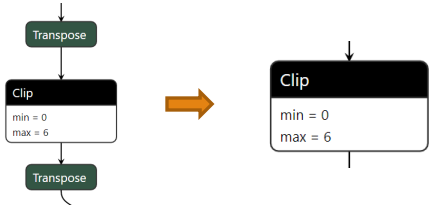

Conversion Process

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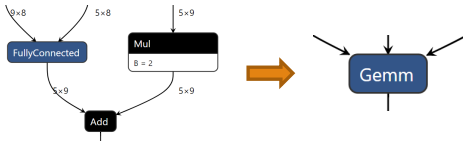

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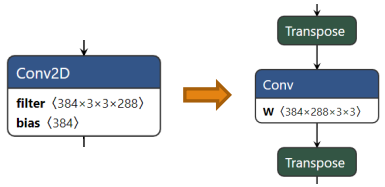

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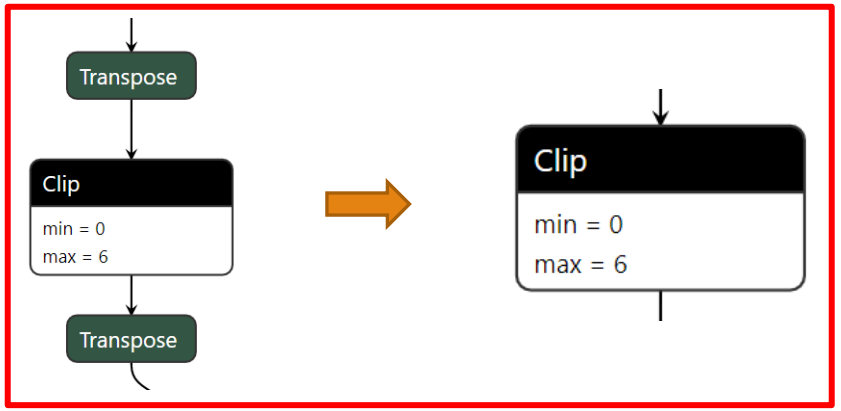
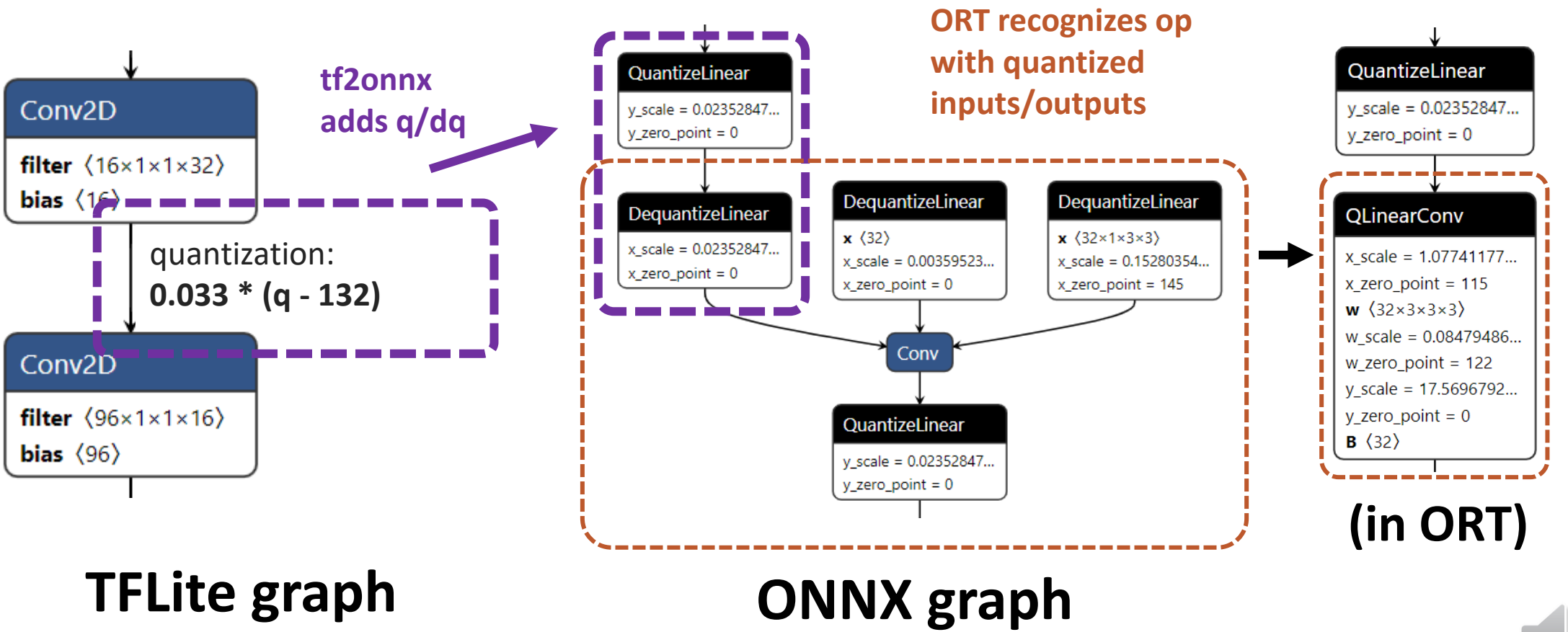
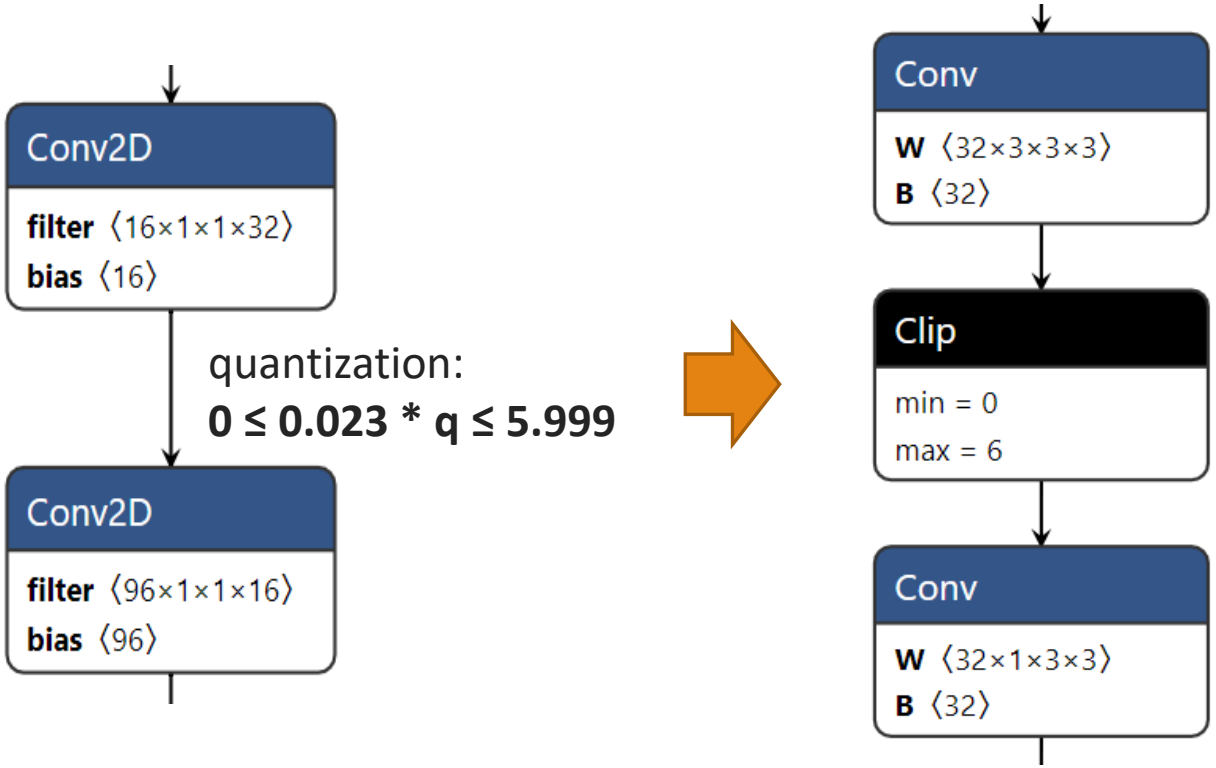


Quantization



Dequantizing Models `--dequantize`

Detect ReLU and ReLU6 ops from quantization range



Support and Feature Requests

Please submit feature requests to GitHub

TFLite -> ONNX conversion is new, expect improvements as we support more ops

github.com/onnx/tensorflow-onnx

ONNX Runtime Mobile

SCOTT MCKAY
MICROSOFT

ONNX RUNTIME MOBILE TECHNICAL LEAD

ONNX Runtime Mobile

ONNX Runtime Mobile is a variant of ONNX Runtime that minimizes binary size for mobile and edge scenarios

- Same codebase as ONNX Runtime
- Available since ONNX Runtime v1.5, Sept 2020

Includes only required operator kernels in the build

- Can also reduce types supported by operator kernels

Custom format for the model file

ONNX Runtime Mobile

Runtime usage of ONNX Runtime Mobile is the same as regular ONNX Runtime

- C, C++, Python and Java APIs are available

Supports NNAPI Execution Provider on Android

Supports CoreML Execution Provider on iOS (preview)

Documentation:

- ONNX_Runtime_for_Mobile_Platforms.md

ORT format model

Created from an ONNX model

- Python script handles conversion

During conversion:

- ONNX Runtime optimizations are applied
  - e.g. constant folding
- Nodes are assigned to kernels
  - No ONNX schema dependency
    - Significant binary size and memory usage saving

Uses google::flatbuffers

Operator Kernel selection

Configuration file specifies the kernels to include in the build

- Model conversion script will automatically generate configuration file when converting models
- Configuration file can also be manually created/edited

Example config:

- `ai.onnx;11;AveragePool,Conv,Reshape,Shape,Softmax,Squeeze,Transpose`

Reduced Type Support

Can limit types that operator kernels support

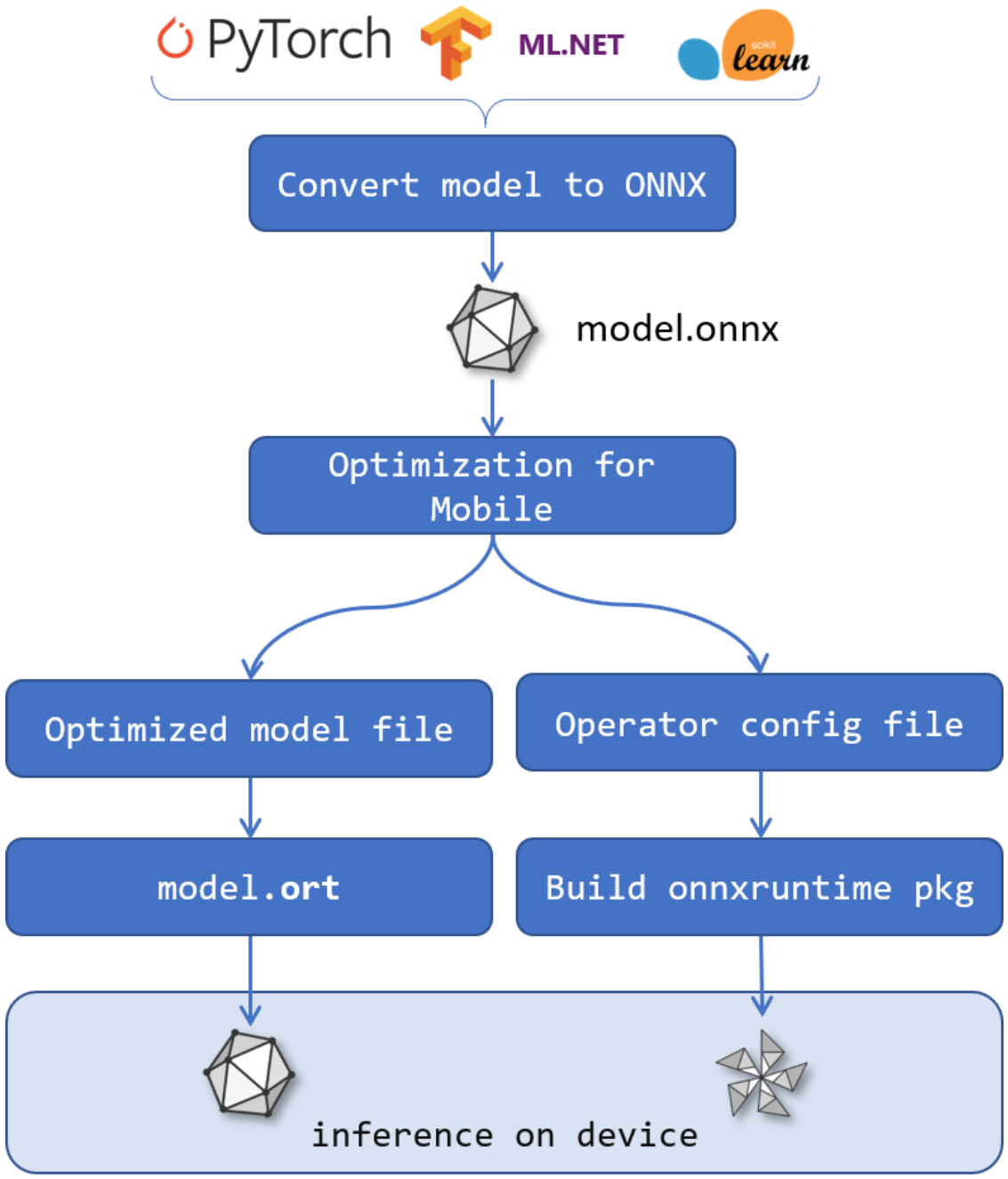
- Model conversion script can automatically detect required types on a per-operator basis
- Alternatively, can specify a global list of types to support

Model based type reduction generally reduces kernel binary size by 25 - 33%

Available in ONNX Runtime v1.7

- March 2021

ORT Mobile Usage



Binary size

Primary choices that determine binary size:

- Operators and types to include
- Enable/disable exceptions
- Enable/disable support for traditional ML operators
- Use static or shared libc++ on Android

Base build size for Android ARM64 *NDK 21.1, no operator kernels, shared libc++, exceptions and traditional ML support disabled*	libonnxruntime.so: 755KB (280KB in AAR)
With operator kernels required by Mobilenet	libonnxruntime.so: 895KB (342KB in AAR)
With reduced type support enabled	libonnxruntime.so: 851KB (325KB in AAR) 31% reduction in size of kernels

NNAPI Support

Usage of NNAPI is determined at runtime

- based on whether NNAPI is available and device capabilities
  - e.g. older version of NNAPI may not support as many operators

Fallback to CPU execution if node cannot be run using NNAPI

Available in ORT v1.6

- December 2020

Questions and Feature Requests

Please reach out to the ONNX Runtime team

- https://github.com/microsoft/onnxruntime/discussions