

Integrate Arm NN into an Android app

Version 21.11

Tutorial

Non-Confidential

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Tutorial

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Release information

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1 Introduction

1.1 Conventions

The following subsections describe conventions used in Arm documents.

Glossary

The Arm Glossary is a list of terms used in Arm documentation, together with definitions for those terms. The Arm Glossary does not contain terms that are industry standard unless the Arm meaning differs from the generally accepted meaning.

See the Arm® Glossary for more information: developer.arm.com/glossary.

Typographic conventions

Arm documentation uses typographical conventions to convey specific meaning.

Convention	Use Use
italic	Introduces special terminology, denotes cross-references, and citations.
bold	Highlights interface elements, such as menu names. Denotes signal names. Also used for terms in descriptive lists, where appropriate.
monospace	Denotes text that you can enter at the keyboard, such as commands, file and program names, and source code.
monospace italic	Denotes arguments to monospace text where the argument is to be replaced by a specific value.
monospace bold	Denotes language keywords when used outside example code.
monospace <u>underline</u>	Denotes a permitted abbreviation for a command or option. You can enter the underlined text instead of the full command or option name.
<and></and>	Encloses replaceable terms for assembler syntax where they appear in code or code fragments. For example:
	MRC p15, 0, <rd>, <crn>, <opcode_2></opcode_2></crn></rd>
SMALL CAPITALS	Used in body text for a few terms that have specific technical meanings, that are defined in the Arm Glossary. For example, IMPLEMENTATION DEFINED, IMPLEMENTATION SPECIFIC, UNKNOWN, and UNPREDICTABLE.
Caution	This represents a recommendation which, if not followed, might lead to system failure or damage.
Warning	This represents a requirement for the system that, if not followed, might result in system failure or damage.
Danger	This represents a requirement for the system that, if not followed, will result in system failure or damage.

Convention	Use
Note	This represents an important piece of information that needs your attention.
- Tip	This represents a useful tip that might make it easier, better or faster to perform a task.
Remember	This is a reminder of something important that relates to the information you are reading.

1.2 Additional reading

This document contains information that is specific to this product. See the following documents for other relevant information:

Table 1-2: Arm publications

Document Name	Document ID	Licensee only
None	-	-

1.3 Feedback

Arm welcomes feedback on this product and its documentation.

Feedback on this product

If you have any comments or suggestions about this product, contact your supplier and give:

- The product name.
- The product revision or version.
- An explanation with as much information as you can provide. Include symptoms and diagnostic procedures if appropriate.

Feedback on content

If you have comments on content then send an e-mail to errata@arm.com. Give:

- The title Integrate Arm NN into an Android app Tutorial.
- The number 102744 2111 01 en.
- If applicable, the page number(s) to which your comments refer.
- A concise explanation of your comments.

Arm also welcomes general suggestions for additions and improvements.



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1.4 Other information

See the Arm website for other relevant information.

- Arm® Developer.
- Arm® Documentation.
- Technical Support.
- Arm® Glossary.

2 Overview

This guide provides instructions on how to use the Arm NN TfLite Delegate in an Android appusing Android Studio.

The guide shows you how easy it is to integrate Arm NN into an existing app that uses the TfLite Interpreter.

Before you begin

This guide uses Android Studio to modify an existing Android app. Therefore, you must have Android Studio installed. You can find installation instructions on the Android-Webpage.

As a basis this guide uses the Image-Segmentation example applications from the Tensorflow-Examples repository. Do not forget to clone the Tensorflow-Examples repository onto your machine using git.

To try out the example yourself, you must have an Android device with a Cortex-A CPU or Mali GPU. Otherwise, you will not be able to benefit from the hardware acceleration that Arm NN offers. If you are not sure if your device has a matching CPU or GPU, you can use the utility functions in the Integrate Arm NN in Android app section to test if Arm NN is supported on your device.

3 Integrate Arm NN in Android app

This section shows you how to change the Image-Segmentation example to use Arm NN to run the neural network in the app.

Introduction

For an introduction to what the Image-Segmentation app does and what its features are, read the read-me file in the examples directory examples/lite/examples/image_segmentation/android/README.md.

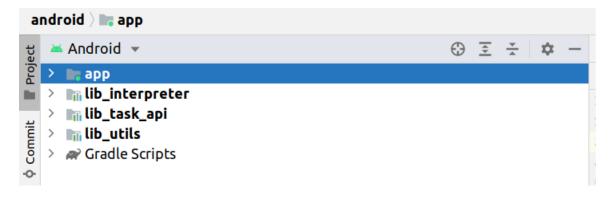
Arm NN provides an Android Library (AAR) that gives you access to the Arm NN TfLite Delegate. This library can be added to any Android app as a dependency. Once you have created an Arm NN Delegate, you can pass it on to the TfLite interpreter which uses the Arm NN delegate to execute models. Before you try that out, you must load the example app into Android Studio and test the original app. The procedure is described in the following section.

Test the original app

Open the image-segmentation example from the Tensorflow-Examples repository with Android Studio File > Open > examples/lite/examples/image_segmentation/android.

The initial setup might take a moment but once that is done you will see the following structure in your Project view.

Figure 3-1: Project structure of the Image-Segmentation app



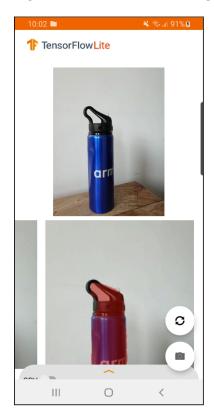
The app offers two different ways to do image segmentation, one is using the Task-API and the other is using the TfLite-Interpreter. This guide changes the interpreter variant of the app to use Arm NN. To change to the interpreter build variant go to Build > select Build Variant and select interpreterDebug.

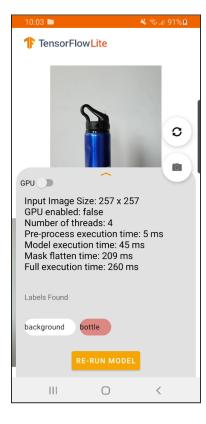
Once the correct build variant is selected, it is time for a test run. Connect a phone to your machine. The device must be in developer mode to be used in Android Studio. You can find more instructions on how to run apps on hardware devices here.

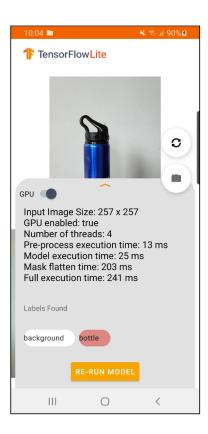
The following figure shows the app having run correctly. In the following figure:

- 1. The left image shows a screen-shot of the app just after an image has been taken. The area in which the bottle was detected is marked.
- 2. The middle image shows execution times using the CPU of the phone.
- 3. The right image shows execution times when enabling the GPU using the switch button.

Figure 3-2: Test run of the original Image-Segmentation app.







You can use the switch button GPU to enable the Tensorflow Gpu-Delegate which runs the underlying neural network on the GPU. Pay attention to the reported model execution times for a later comparison.

If the app runs correctly, you can proceed to the next step.

Modify the app

The steps in this section cover modifying the app to use Arm NN. The goal is to change the switch button to enable Arm NN instead of the GPU.

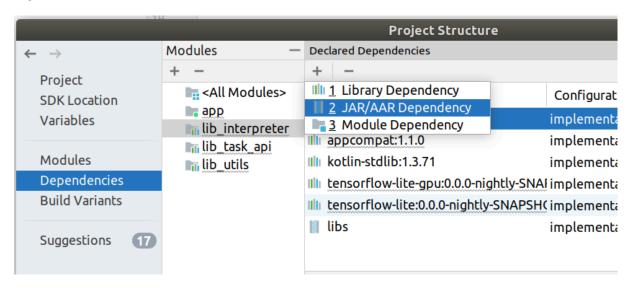
Procedure

1. Download the Arm NN Android Library from our GitHub release page. It is uploaded as an asset to the latest release.

2. Add the Arm NN module as a dependency to the lib_interpreter module.

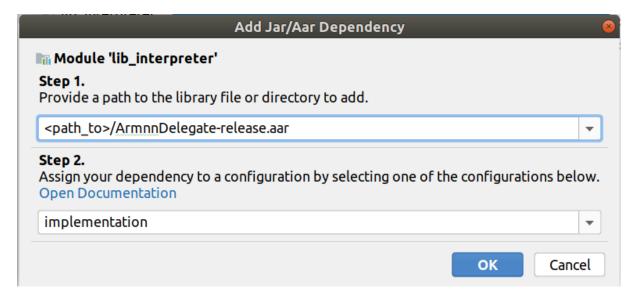
Open the Project Structure view File > Project Structure. Choose Dependencies then the lib_interpreter module and after that click the + button to add a JAR/AAR Dependency.

Figure 3-3: Add the Arm NN Library as dependency



In the next dialog, add the directory path to the Arm NN Android Library that you have downloaded in the first step.

Figure 3-4: Add the Arm NN Library as dependency

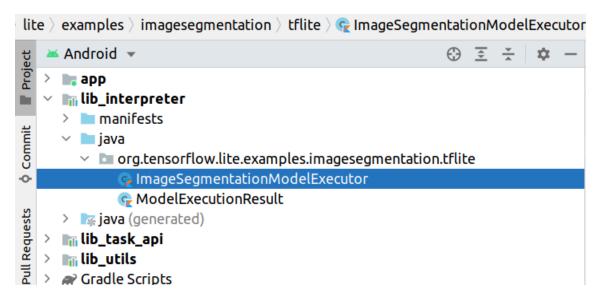


3. Modify the lib_interpreter module.

First we update the kotlin-gradle-plugin version to make the example compatible with the Arm NN module. To do so, navigate to the image-segmentation/android/build.gradle file and change the line marked with - to the line marked with +:

In the next step, the <code>lib_interpreter</code> module is changed to use Arm NN when the switch button is enabled. This follows the same procedure as the TfLite GPU delegate. Open the <code>ImageSegmentationModelExecutor.kt</code> file in the <code>lib</code> interpreter module.

Figure 3-5: Open ImageSementationModelExecutor.kt



And then apply the following changes:

- Change the Boolean variable useGPU to useArmnn.
- Initialize a variable to hold an instance of the ArmnnDelegate.
- In the getInterpreter function, rename the parameter useGPU to useArmNN.
- When the TfLite interpreter is created, create an Arm NN delegate if requested. Additional options can be given to the ArmnnDelegate in the form of two arrays of strings. The first contains the name of the options and the second contains the value the options should be changed to. In this example, logging is turned on and set to info. Arm NN supports multiple backends. To run a model on Arm Cortex-A CPUs, set the backend option to

cpuace. However, if you want to run the model on Arm Mali GPUs use Gpuace. There are multiple options for optimization purposes that can be configured to increase the execution speed. You can find more information on these in the Arm NN module description or in our documentation on GitHub.

- Before you attempt to execute a model with cpuace or gpuace, you must confirm that the device supports that backend. This can be done with the ArmnnUtils functions supplied with the library.
- Because the TfLites GPU delegate is usually faster than the CPU, we change the app to use the GPU delegate by default. That means if the switch is turned off, the GPU delegate is used and if it is turned on the Arm NN delegate is used.
- The last code change in this file is to destroy the Arm NN delegate when the close () function is called.

The changes that were described can be seen in the following code:

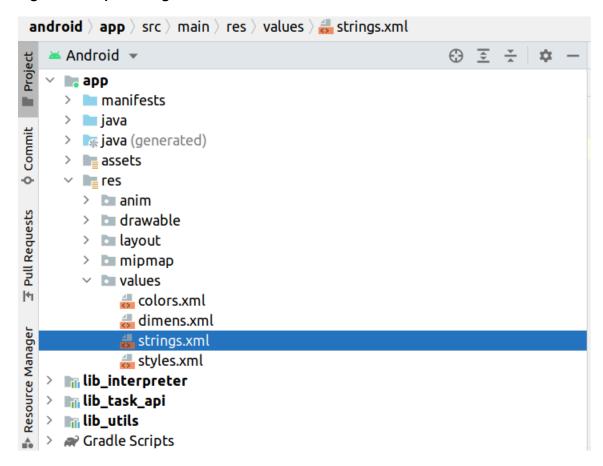
```
@@ -20,10 +20,12 @@ import android.content.Context
 import android.graphics.Bitmap
 import android.graphics.Color
 import android.os.SystemClock
 import androidx.core.graphics.ColorUtils
 import android.util.Log
+import com.arm.armnn.delegate.ArmnnDelegate
+import com.arm.armnn.delegate.ArmnnUtils
 import java.io.FileInputStream
 import java.io.IOException
import java.nio.ByteBuffer
 import java.nio.ByteOrder
 import java.nio.MappedByteBuffer
@@ -45,13 +47,14 @@ import org.tensorflow.lite.gpu.GpuDelegate
  * 'car', 'cat', 'chair', 'cow', 'diningtable', 'dog', 'horse', 'motorbike', 
* 'person', 'pottedplant', 'sheep', 'sofa', 'train', 'tv'
 class ImageSegmentationModelExecutor(
   context: Context,
  private var useGPU: Boolean = false
   private var useArmNN: Boolean = false
   private var gpuDelegate: GpuDelegate? = null
  private var armnnDelegate: ArmnnDelegate? = null
   private val segmentationMasks: ByteBuffer
   private val interpreter: Interpreter
   private var fullTimeExecutionTime = 0L
@@ -61,11 +64,11 @@ class ImageSegmentationModelExecutor(
   private var numberThreads = 4
   init {
     interpreter = getInterpreter(context, imageSegmentationModel, useGPU)
     interpreter = getInterpreter(context, imageSegmentationModel, useArmNN)
segmentationMasks = ByteBuffer.allocateDirect(1 * imageSize * imageSize *
 NUM CLASSES * 4)
     segmentationMasks.order(ByteOrder.nativeOrder())
   fun execute(data: Bitmap): ModelExecutionResult {
@@ -147,28 +150,41 @@ class ImageSegmentationModelExecutor(
   @Throws(IOException::class)
   private fun getInterpreter(
```

```
context: Context,
     modelName: String,
useGpu: Boolean = false
     useArmNN: Boolean = false
   ): Interpreter {
     val tfliteOptions = Interpreter.Options()
     tfliteOptions.setNumThreads(numberThreads)
     gpuDelegate = null
     if (useGpu) {
     armnnDelegate = null
     if (useArmNN) {
+
       if (ArmnnUtils.isGpuAccBackendSupportedOnThisDevice()) {
+ var optionKeys = arrayOf("logging-severity", "backends", "enable-fast-math", "reduce-fp32-to-fp16", "memory-import")
         var optionValues = arrayOf(
                                                    "info",
                                                               "GpuAcc",
 "true",
                          "true",
                                            "true")
         armnnDelegate = ArmnnDelegate(optionKeys, optionValues)
         tfliteOptions.addDelegate(armnnDelegate)
+
       else {
         Log.w(TAG, "The Arm NN GpuAcc backend is not supported on this device.")
+
     // the gpu delegate is enabled by default
+
       gpuDelegate = GpuDelegate()
        tfliteOptions.addDelegate(gpuDelegate)
     return Interpreter (loadModelFile (context, modelName), tfliteOptions)
   private fun formatExecutionLog(): String {
     val sb = StringBuilder()
     sb.append("Input Image Size: $imageSize x $imageSize\n")
     sb.append("GPU enabled: $useGPU\n") sb.append("ArmNN enabled: $useArmNN\n")
     sb.append("Number of threads: $numberThreads\n")
     sb.append("Pre-process execution time: $preprocessTime ms\n")
     sb.append("Model execution time: $imageSegmentationTime ms\n")
     sb.append("Mask flatten time: $maskFlatteningTime ms\n")
     sb.append("Full execution time: $fullTimeExecutionTime ms\n")
@@ -178,10 +194,13 @@ class ImageSegmentationModelExecutor(
   fun close() {
     interpreter.close()
     if (gpuDelegate != null) {
       gpuDelegate!!.close()
     if (armnnDelegate != null) {
       armnnDelegate!!.close()
   private fun convertBytebufferMaskToBitmap(
     inputBuffer: ByteBuffer,
```

```
imageWidth: Int,
```

• To make it clearer which delegate is being used, we can change the label of the switch button from GPU to Armnn. This can be done with the following change in image-segmentation/android/app/src/main/res/values/strings.xml

Figure 3-6: Open string.xml



Test Arm NN

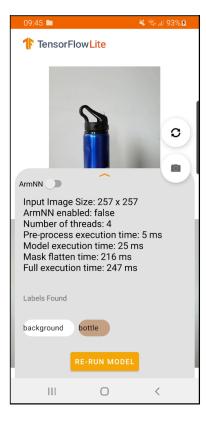
Now that all changes to the app have been made, we can try out the app.

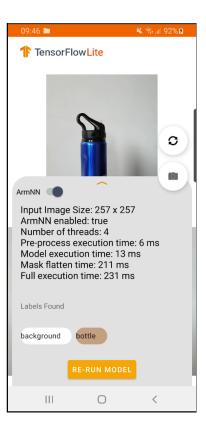
As you can see in the following figure, the performance has improved significantly when using the Arm NN delegate. In the following figure:

- 1. The left image shows a screen-shot of the app just after an image has been taken. At this stage, the app uses the TfLite GPU delegate.
- 2. The middle image shows the performance figures with the TfLite GPU delegate.
- 3. The right images shows the performance using the Arm NN delegate.

Figure 3-7: Test run of the Image-Segmentation app with Arm NN.







When you try this out yourself, you will notice quite a bit of noise in the model execution time regardless of which backend is applied. When taking the images, we have tried to represent the average execution time we have been seeing on our test device.

These performance numbers are estimates. Even though Arm NN usually turns out to be faster, it cannot be generalized that Arm NN is always faster than the TfLite GPU delegate. The resulting performance depends on the model and the operations it contains. Models can also be optimized to run on specific hardware. The hardware in the device also plays a significant role.

4 Tips and Tricks

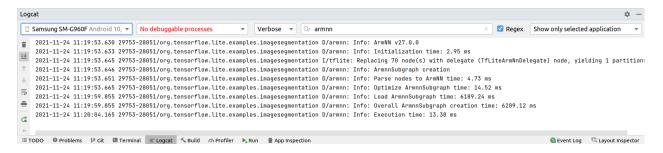
This section contains tips to help you if you have trouble with the Arm NN Android Library.

Enable logging

If you cannot verify that Arm NN is running after you have integrated the Arm NN delegate, look at the logging output in logcat. You can view the logging output in Android Studio.

Apply a filter armnn to the logical output while running the app. If you have set the logging-severity option of the Arm NN Delegate to info, as shown in this guide, you will see an output similar to the following image.

Figure 4-1: Arm NN logcat output



You can set the <code>logging-severity</code> level to <code>debug</code> to get even more information about which optimization options have been used. If you want no logging at all, do not provide <code>logging-severity</code> to the options.

5 Related information

Here are some resources related to the material in this guide:

- Arm Community ask development questions and find articles and blogs on specific topics from Arm experts.
- Arm NN Github raise queries or issues associated with the Arm NN how-to guides.
- Arm NN Product Documentation find out more about the latest Arm NN features.

Appendix A Revisions

This appendix describes the technical changes between released issues of this book.

Table A-1: First release for version 21.11

Change	Location
First release	_