

Running GGUF Models Offline on Device in Flutter (Tier-1 & Tier-2)

Overview

This guide explains how to run your fine-tuned Qwen-2.5 GGUF models **locally and offline** inside a Flutter app. The architecture uses two separate models:

- **Tier-1 (Router):** A small, always-on model (Qwen-2.5-1.5B) that classifies intent, routes requests, and outputs strict JSON. It never generates long text.
- **Tier-2 (Reasoner):** A larger, on-demand model (Qwen-2.5-3B or 7B) that handles deep reasoning, explanations, and planning. It outputs natural language.

Both models are exported as `.gguf` files after fine-tuning and run on-device using a C++ inference engine (llama.cpp) via Flutter's FFI bridge.

Understanding GGUF File Formats: F16 vs Q4_K_M

After fine-tuning, you export your model to GGUF. The two most common formats are **F16** (full precision) and **Q4_K_M** (quantized). Understanding the difference is critical for choosing the right file for each tier.

F16 (Float16 / Half Precision)

- **What it is:** The model weights are stored in 16-bit floating point. This is essentially the full-quality model with no compression.
- **File size:** Large. A 1.5B parameter model is ~3 GB; a 3B model is ~6 GB; a 7B model is ~14 GB.
- **Quality:** Maximum accuracy. No quality loss from the trained weights.
- **Speed:** Slower inference on mobile because more data must be read from memory per token.
- **RAM usage:** High. The entire model must fit in RAM at full precision.
- **When to use F16:**
 - During development and testing on a desktop/emulator to validate model correctness.
 - As the source file for producing quantized versions (you quantize FROM F16).
 - On high-end devices with abundant RAM where you want maximum quality and speed is not a concern.
 - **Never ship F16 to production mobile apps** unless targeting tablets or very high-end devices with 12+ GB RAM.

Q4_K_M (4-bit Quantization, K-Quant Medium)

- **What it is:** The model weights are compressed to ~4.5 bits per weight using the K-quant method with medium quality settings. This is a lossy compression, but the quality loss is minimal for well-trained models.
- **File size:** Much smaller. A 1.5B parameter model is ~900 MB; a 3B model is ~1.8 GB; a 7B model is ~4 GB.
- **Quality:** Very close to F16 for most tasks. Routing/classification (Tier-1) and natural language generation (Tier-2) both survive Q4_K_M quantization well. Minor degradation is possible on edge-case reasoning tasks.
- **Speed:** Faster inference on mobile because less memory bandwidth is needed per token.
- **RAM usage:** Significantly lower. This is what makes on-device inference practical.
- **When to use Q4_K_M:**
 - **Production mobile apps** – this is the standard format for shipping.
 - Both Tier-1 and Tier-2 models in your Flutter app.
 - Any device with limited RAM (which is every phone).

Quick Comparison Table

Aspect	F16 (Half Precision)	Q4_K_M (4-bit Quantized)
Bits per weight	16	~4.5
1.5B model size	~3 GB	~900 MB
3B model size	~6 GB	~1.8 GB
7B model size	~14 GB	~4 GB
Quality loss	None	Minimal
Inference speed	Slower	Faster
RAM required	High	Low
Use case	Dev/testing, source for quantization	Production mobile deployment

How to Convert F16 to Q4_K_M

If you have an F16 GGUF and need to quantize it:

```
# Using llama.cpp's quantize tool
./quantize model_f16.gguf model_q4_k_m.gguf Q4_K_M
```

You always fine-tune first, export to F16 GGUF, then quantize to Q4_K_M for deployment.

Which GGUF Files You Need

For the Tier-1 / Tier-2 architecture, you will have **two separate GGUF files**:

File	Model	Role	Recommended Format	Approximate Size
tier1_router_q4_k_m.gguf	Qwen-2.5-1.5B (fine-tuned)	Intent routing, JSON output	Q4_K_M	~900 MB
tier2_reasoner_q4_k_m.gguf	Qwen-2.5-3B (fine-tuned)	Reasoning, natural language	Q4_K_M	~1.8 GB

Total on-device storage: ~2.7 GB (both models combined with Q4_K_M).

Note: You do NOT need to load both models into RAM simultaneously. Tier-1 stays loaded. Tier-2 is loaded on-demand and unloaded when not needed.

Step 1: Add Dependencies

Update your `pubspec.yaml`:

```
dependencies:
  flutter:
    sdk: flutter
  llama_cpp_dart: ^0.1.3      # C++ inference engine FFI bindings
  path_provider: ^2.1.2      # Access to device file system paths

flutter:
  assets:
    - assets/models/tier1_router_q4_k_m.gguf
    # Tier-2 is NOT bundled as an asset (too large).
    # It is downloaded separately after install. See Step 4.
```

Why Tier-1 is bundled but Tier-2 is not

- Tier-1 (~900 MB):** Small enough to bundle inside the APK/IPA. The app works immediately after install because the router is always available.
- Tier-2 (~1.8 GB+):** Too large to bundle. It would bloat the app download beyond acceptable limits. Instead, download it on first launch (or on-demand) to the device's local storage.

If your Tier-1 model is also too large to bundle (e.g., you used a bigger base model), you can download both models post-install and skip the `assets` declaration entirely.

Step 2: Configure Native Platforms

Android

android/app/src/main/AndroidManifest.xml:

```
<application
  android:largeHeap="true"
  android:requestLegacyExternalStorage="true">
  <!-- android:largeHeap gives the app access to more RAM -->
  <!-- Required for loading GGUF models into memory -->
</application>
```

android/app/build.gradle:

```

android {
  defaultConfig {
    // Set minimum SDK to 24 (Android 7.0) for 64-bit support
    minSdkVersion 24
  }

  // Prevent compression of GGUF files in the APK
  aaptOptions {
    noCompress 'gguf'
  }
}

```

Important: Older Android devices (pre-Android 10) have a **4 GB file size limit** for individual assets. Q4_K_M for 1.5B and 3B models are both under this limit.

iOS

ios/Runner/Info.plist:

- Set deployment target to **iOS 14.0+**
- No additional configuration needed for file access since `getApplicationDocumentsDirectory()` is within the app sandbox.

ios/Podfile:

```
platform :ios, '14.0'
```

Step 3: Implement Model Services

You need two service classes – one for each tier. Both use the same llama.cpp engine but with different models and different usage patterns.

Tier-1 Model Service (Always-On Router)

Create `lib/services/tier1_service.dart`:

```

import 'dart:io';
import 'dart:convert';
import 'package:flutter/services.dart';
import 'package:path_provider/path_provider.dart';
import 'package:llama_cpp_dart/llama_cpp_dart.dart';

class Tier1Service {
  LlamaProcessor? _processor;
  bool _isloaded = false;

  /// Copies the Tier-1 model from app assets to local storage.
  /// The C++ engine needs a real file path -- it cannot read from
  /// Flutter's asset bundle directly.
  Future<String> _copyAssetToLocal() async {
    final dir = await getApplicationDocumentsDirectory();
    final filePath = '${dir.path}/tier1_router_q4_k_m.gguf';
    final file = File(filePath);

    if (!await file.exists()) {
      final byteData = await rootBundle.load(
        'assets/models/tier1_router_q4_k_m.gguf',
      );
      await file.writeAsBytes(
        byteData.buffer.asUint8List(
          byteData.offsetInBytes,
          byteData.lengthInBytes,
        ),
      );
    }
    return filePath;
  }
}

```

```

/// Load Tier-1 into memory. Call this once at app startup.
/// Tier-1 stays loaded for the entire app session.
Future<void> load() async {
  if (!_isLoading) return;

  final modelPath = await _copyAssetToLocal();
  final params = ModelParams()
    ..nCtx = 512      // Small context window (routing needs little)
    ..nThreads = 4;   // Adjust based on device CPU cores

  _processor = LlamaProcessor(path: modelPath, modelParams: params);
  _processor?.load();
  _isLoading = true;
}

/// Run a user message through Tier-1 and parse the JSON response.
/// Returns a structured routing decision.
Future<Map<String, dynamic>> route(String userMessage) async {
  if (!_isLoading) await load();

  final systemPrompt = 'You are a Tier-1 router. Output JSON only.';
  final prompt = '<|im_start|>system\n$systemPrompt<|im_end|>\n'
    '<|im_start|>user\n$userMessage<|im_end|>\n'
    '<|im_start|>assistant\n';

  final buffer = StringBuffer();
  await for (final token in _processor!.stream(prompt)) {
    buffer.write(token);
  }

  return jsonDecode(buffer.toString()) as Map<String, dynamic>;
}

void dispose() {
  _processor?.unload();
  _isLoading = false;
}
}

```

Key points about Tier-1:

- Loaded once at app startup, stays in memory the entire session.
- Uses a small context window (nCtx = 512) because routing decisions are short.
- Output is always strict JSON matching the Tier-1 schema (intent, journey, tool, complexity_score, etc.).
- Uses **ChatML format** (<|im_start|> / <|im_end|> tokens) because the Qwen-2.5 models are pretrained with this format.

Tier-2 Model Service (On-Demand Reasoner)

Create `lib/services/tier2_service.dart`:

```

import 'dart:io';
import 'package:path_provider/path_provider.dart';
import 'package:llama_cpp_dart/llama_cpp_dart.dart';

class Tier2Service {
  LlamaProcessor? _processor;
  bool _isLoading = false;
  bool _isDownloaded = false;

  /// Check if the Tier-2 model has been downloaded to local storage.
  Future<bool> isAvailable() async {
    final dir = await getApplicationDocumentsDirectory();
    final file = File('${dir.path}/tier2_reasoner_q4_k_m.gguf');
    _isDownloaded = await file.exists();
  }
}

```

```

    return _isDownloaded;
}

/// Download the Tier-2 model from your server.
/// Call this on first launch or when the user opts in.
/// Returns a stream of download progress (0.0 to 1.0).
Stream<double> download(String downloadUrl) async* {
    final dir = await getApplicationDocumentsDirectory();
    final filePath = '${dir.path}/tier2_reasoner_q4_k_m.gguf';
    final file = File(filePath);

    // Use your preferred HTTP client (dio, http, etc.)
    // This is pseudocode -- replace with your actual download logic.
    final httpClient = HttpClient();
    final request = await httpClient.getUrl(Uri.parse(downloadUrl));
    final response = await request.close();
    final totalBytes = response.contentLength;
    var receivedBytes = 0;

    final sink = file.openWrite();
    await for (final chunk in response) {
        sink.add(chunk);
        receivedBytes += chunk.length;
        if (totalBytes > 0) {
            yield receivedBytes / totalBytes;
        }
    }
    await sink.close();
    httpClient.close();
    _isDownloaded = true;
}

/// Load Tier-2 into memory. Only call when needed.
Future<void> ensureLoaded() async {
    if (_isLoading) return;
    if (!_isDownloaded) {
        final available = await isAvailable();
        if (!available) {
            throw StateError('Tier-2 model not downloaded yet');
        }
    }

    final dir = await getApplicationDocumentsDirectory();
    final modelPath = '${dir.path}/tier2_reasoner_q4_k_m.gguf';

    final params = ModelParams()
        ..nCtx = 2048 // Larger context for reasoning tasks
        ..nThreads = 4;

    _processor = LlamaProcessor(path: modelPath, modelParams: params);
    _processor?.load();
    _isLoading = true;
}

/// Generate a reasoning response. Returns a stream of tokens
/// so you can display the response as it generates (streaming UX).
Stream<String> generate(String userMessage) {
    if (!_isLoading) {
        return Stream.value('Tier-2 model is not loaded.');
```

```

    }

    final systemPrompt = 'You are a helpful reasoning assistant.';
    final prompt = '<|im_start|>system\n$systemPrompt<|im_end|>\n'
        '<|im_start|>user\n$userMessage<|im_end|>\n'
        '<|im_start|>assistant\n';

```

```

        return _processor!.stream(prompt);
    }

    /// Unload Tier-2 from memory to free RAM.
    /// Call this when reasoning is complete or on memory pressure.
    void unload() {
        _processor?.unload();
        _processor = null;
        _isLoading = false;
    }

    void dispose() => unload();
}

```

Key points about Tier-2:

- **Not bundled** with the app. Downloaded separately to device storage.
- **Loaded on-demand** only when Tier-1's `complexity_score` exceeds the threshold (e.g., > 60).
- **Unloaded after use** to free RAM for the rest of the app.
- Uses a larger context window (`nCtx` = 2048) because reasoning tasks produce longer outputs.
- Output is natural language, not JSON.

Step 4: Implement the Orchestrator

The orchestrator ties Tier-1 and Tier-2 together. This is the core control logic of your app.

Create `lib/services/orchestrator.dart` :

```

import 'tier1_service.dart';
import 'tier2_service.dart';

class Orchestrator {
    final Tier1Service _tier1 = Tier1Service();
    final Tier2Service _tier2 = Tier2Service();

    static const int complexityThreshold = 60;

    /// Initialize -- load Tier-1 at startup. Tier-2 stays dormant.
    Future<void> init() async {
        await _tier1.load();
    }

    /// Main entry point: process any user message.
    Future<OrchestratorResult> handle(String userMessage) async {
        // Step 1: Always run Tier-1 first
        final routing = await _tier1.route(userMessage);

        // Step 2: If Tier-1 needs clarification, ask the user
        if (routing['needs_clarification'] == true) {
            return OrchestratorResult(
                type: ResultType.clarification,
                text: routing['clarification'] as String,
            );
        }

        // Step 3: If complexity is high, escalate to Tier-2
        final complexity = routing['complexity_score'] as int? ?? 0;
        if (complexity > complexityThreshold) {
            await _tier2.ensureLoaded();

            final buffer = StringBuffer();
            await for (final token in _tier2.generate(userMessage)) {
                buffer.write(token);
            }
        }
    }
}

```

```
// Optionally unload Tier-2 to free memory
_tier2.unload();

return OrchestratorResult(
    type: ResultType.reasoning,
    text: buffer.toString(),
    routing: routing,
);
}

// Step 4: If a cloud tool is needed, call it
if (routing['needs_tools'] == true) {
    return OrchestratorResult(
        type: ResultType.toolCall,
        text: null,
        routing: routing,
        // Flutter will now call the cloud API using
        // routing['tool'] and routing['arguments']
    );
}

// Step 5: Simple response -- Tier-1 handled it entirely
return OrchestratorResult(
    type: ResultType.local,
    routing: routing,
);
}

void dispose() {
    _tier1.dispose();
    _tier2.dispose();
}
}

enum ResultType { clarification, reasoning, toolCall, local }

class OrchestratorResult {
    final ResultType type;
    final String? text;
    final Map<String, dynamic>? routing;

    OrchestratorResult({required this.type, this.text, this.routing});
}
```

Step 5: Memory Management Strategy

Running models on-device requires careful memory management. Here is how to handle both tiers without crashing the app.

RAM Budget by Device Tier

Device Class	Available RAM	Strategy
Low-end (3-4 GB RAM)	~1.5 GB for app	Tier-1 only (Q4_K_M 1.5B). Skip Tier-2.
Mid-range (6-8 GB RAM)	~3 GB for app	Tier-1 always loaded + Tier-2 (3B Q4_K_M) on-demand. Unload Tier-2 after each use.
High-end (8-12+ GB RAM)	~5 GB for app	Tier-1 always loaded + Tier-2 (3B or 7B Q4_K_M) can stay resident longer.

Memory Rules

- 1. **Tier-1 stays loaded** for the entire session. At ~900 MB (Q4_K_M 1.5B), this is manageable on all modern phones.
- 2. **Tier-2 loads only when `complexity_score > threshold`** and unloads immediately after the response is complete.
- 3. **Monitor memory pressure.** On Android, listen to `ComponentCallbacks2.onTrimMemory()` . On iOS, respond to `UIApplication.didReceiveMemoryWarningNotification` . If triggered, unload Tier-2 immediately.
- 4. **Never load both models at full context simultaneously** on low-end devices. If Tier-2 needs to load, ensure Tier-1's context is minimal (it should be, since routing is already done).

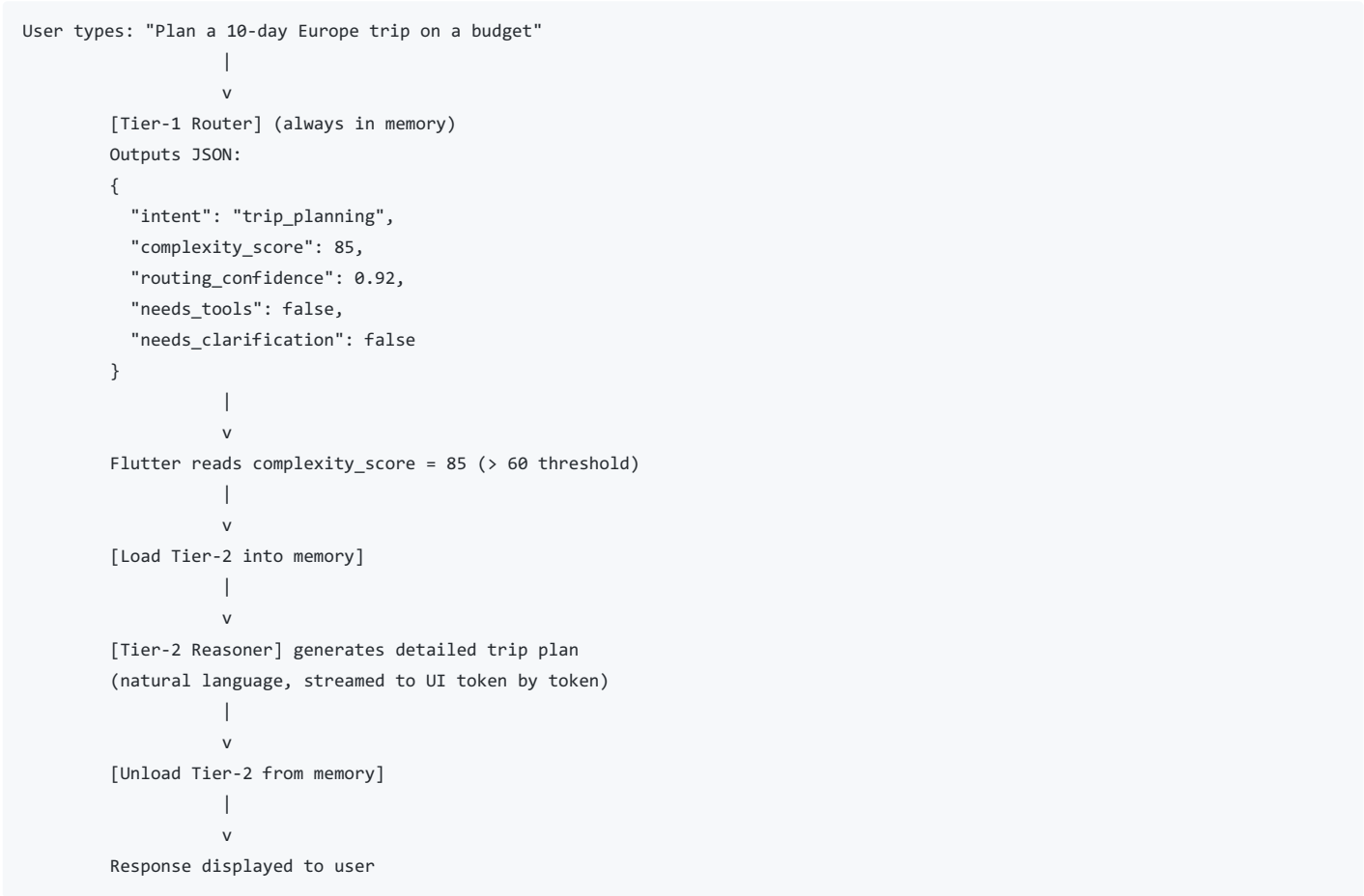
Step 6: File Structure

Here is the recommended project structure for the model-related files:

```
lib/
  services/
    tier1_service.dart      # Tier-1 router model service
    tier2_service.dart      # Tier-2 reasoner model service
    orchestrator.dart      # Coordinates Tier-1, Tier-2, and cloud tools
  assets/
    models/
      tier1_router_q4_k_m.gguf # Bundled with the app (if small enough)
```

The Tier-2 model file (`tier2_reasoner_q4_k_m.gguf`) lives in the device's application documents directory after download. It is **not** in the `assets/` folder.

Complete Flow: What Happens When a User Sends a Message





Troubleshooting

Common Issues

Problem	Cause	Fix
App crashes on model load	Not enough RAM	Use Q4_K_M instead of F16. Reduce nctx . Close background apps during testing.
Model outputs garbage	Wrong prompt format	Use ChatML format (< im_start > / < im_end > tokens). Qwen-2.5 requires this.
Tier-1 outputs invalid JSON	Poor fine-tuning data or too aggressive quantization	Validate your dataset. Try Q5_K_M if Q4_K_M causes JSON issues.
Asset copy takes too long	Large model file	Show a loading screen with progress. The copy only happens once (first launch).
Tier-2 download fails	Network issues	Implement retry logic and resume support for large file downloads.
Slow inference	Too many context tokens or too many threads	Reduce nctx . Set nThreads to match physical CPU cores (not logical).
4 GB asset limit on Android	GGUF file exceeds 4 GB	Only affects 7B F16 models. Q4_K_M for 7B is ~4 GB (borderline). Use 3B for safety.

Validating Your Model Works

Before integrating into Flutter, test your GGUF file on desktop using the llama.cpp CLI:

```
# Test Tier-1 (should output JSON)
./main -m tier1_router_q4_k_m.gguf \
  -p "<|im_start|>system\nYou are a Tier-1 router. Output JSON only.<|im_end|>\n<|im_start|>user\nBook a flight to Paris<|im_end|>\n"
  -n 256

# Test Tier-2 (should output natural language)
./main -m tier2_reasoner_q4_k_m.gguf \
  -p "<|im_start|>system\nYou are a helpful reasoning assistant.<|im_end|>\n<|im_start|>user\nExplain quantum computing simply<|im_end|>\n"
  -n 512
```

Summary: Decision Checklist

Decision	Answer
Which format for production?	Q4_K_M for both Tier-1 and Tier-2.
Which format for development?	F16 to validate quality, then quantize.
Bundle Tier-1 in the app?	Yes , if under ~1 GB.
Bundle Tier-2 in the app?	No . Download post-install.
Load both models at once?	No . Tier-1 stays loaded. Tier-2 loads/unloads on demand.
Which prompt format?	ChatML (<code>< im_start ></code> / <code>< im_end ></code>).
Minimum device for Tier-1 only?	4 GB RAM (most modern phones).
Minimum device for Tier-1 + Tier-2?	6 GB RAM (mid-range 2022+ phones).