

Problem Chosen

A

**2026
MCM/ICM
Summary Sheet**

Team Control Number

2611750

River Shield: Mathematical Armor for Water Intake Protection

Summary

summary

Keywords:

keyword1; keyword2; keyword3

Contents

1	Introduction	3
1.1	Problem Background	3
1.2	Problem Restatement	3
1.3	Our Work	3
2	Model Preparation	3
2.1	Assumptions and Justifications	3
2.2	Notations	3
2.3	Data Statement	3
2.4	Our Modeling Philosophy	3
3	Core Model Development	3
3.1	Fundamental Equation and Factor Selection	3
3.1.1	Continuous-Time Battery Dynamics	3
3.1.2	Factor Selection and Simplification Strategy	5
3.1.3	Environmental and Aging Factors	5
3.2	Hardware Components Modeling	5
3.2.1	Display Power Model	5
3.2.2	Processor Power Model (CPU + GPU)	7
3.2.3	Memory and Storage Model	7
3.2.4	Sensor Power Model (GPS, Camera, etc.)	7
3.3	Signaling Modules Modeling	7
3.3.1	Cellular and Wi-Fi Communication	7
3.3.2	Bluetooth and Short-Range Communication	7
3.3.3	Special Note: GPS as Positioning Sensor	7
3.4	Software Role in the Modeling Framework	7
3.4.1	OS Overhead: Embedded in Hardware Parameters	7
3.4.2	Application Impact: Hardware Usage Drivers	7
4	Hardware-Software Parameter Calibration Methodology	7
5	Scenario-Based Analysis and Prediction	7
6	Model Analysis and Evaluation	7
7	AI Use Report	8

1 Introduction

1.1 Problem Background

Smartphones have been

1.2 Problem Restatement

In this problem, we're asked to build a

1.3 Our Work

2 Model Preparation

2.1 Assumptions and Justifications

2.2 Notations

We listed all the parameters and terminologied here. Our model has x parameters in all, involving n depending on the physical configuration and $(x-n)$ tunable according to different scenarios.

We will restate related parameters before the modeling of each components.

2.3 Data Statement

2.4 Our Modeling Philosophy

Our core view is that software applications are not "true consumers" but "managers" of the hardwares which directly consume power. Therefore, building a precise mapping from software operations to hardware components'consumption is essential, worthing a specific "methodology".

Besides, we begin with comprehensive consideration of all possible influencing factors according to [2] and gradually shrinking it by merging, simplifying and justified ignoring.

We also seek to link our model to real-world insights, bridging theory with practice, with the ultimate goal of developing truly applicable forecasting capabilities.

3 Core Model Development

3.1 Fundamental Equation and Factor Selection

Our modeling framework treats smartphone hardware as parallel current-drawing components, yielding a governing differential equation for $SOC(t)$. Software behavior influences battery drain exclusively by modulating hardware operating parameters—screen brightness, CPU utilization, network states, etc. We therefore develop modular sub-models that first translate software activities into hardware settings, then compute the resulting current consumption for each component.

3.1.1 Continuous-Time Battery Dynamics

Based on the definition of State of Charge and the physical meaning of electric current, we obtain the following system of equations:

$$\begin{cases} \text{SOC}(t) = \frac{Q(t)}{Q_{\max}} \\ \frac{dQ}{dt} = -I_{\text{total}}(t) \end{cases} \quad (1)$$

where the first equation defines SOC as the ratio of remaining charge to maximum charge, and the second equation expresses charge conservation during discharge.

Therefore, by differentiating the first equation and substituting the second, we derive the governing differential equation:

$$\frac{d(\text{SOC})}{dt} = -\frac{I_{\text{total}}(t)}{C_{\text{eff}}(T, N_{\text{cycles}})} \quad (2)$$

where:

- $\text{SOC}(t) \in [0, 1]$ is the state of charge at time t (dimensionless)
- $I_{\text{total}}(t)$ is the total discharge current (mA) at time t
- $C_{\text{eff}}(T, N_{\text{cycles}})$ is the effective battery capacity (mAh)

To account for environmental and aging effects that reduce usable battery capacity, we introduce the effective capacity C_{eff} , which is related to the nominal capacity by:

$$C_{\text{eff}} = C_{\text{nominal}} \times f_{\text{temp}}(T) \times f_{\text{aging}}(N_{\text{cycles}}) \quad (3)$$

where:

- C_{nominal} is the manufacturer's rated capacity (e.g., 4000 mAh)
- $f_{\text{temp}}(T)$ is the temperature correction factor ($0.7 \leq f_{\text{temp}} \leq 1.0$)
- $f_{\text{aging}}(N_{\text{cycles}})$ is the aging correction factor based on charge-discharge cycles
- T is the battery temperature ($^{\circ}\text{C}$)
- N_{cycles} is the number of complete charge-discharge cycles

The total discharge current $I_{\text{total}}(t)$ represents the sum of currents drawn by all active smartphone components:

$$I_{\text{total}}(t) = I_{\text{display}}(t) + I_{\text{processor}}(t) + I_{\text{memory}}(t) + I_{\text{network}}(t) + I_{\text{sensors}}(t) \quad (4)$$

Further decomposition and refined modelings are provided in the following subsections.

And each component current incorporates two distinct elements:

- **Baseline overhead:** The minimum power required for the component to remain operational, including essential operating system services and hardware idle states.
- **Software-driven consumption:** Additional power drawn when applications and user activities activate hardware features above their baseline levels.

Time-to-Empty Prediction Given initial conditions $\text{SOC}(t_0) = \text{SOC}_0$, the time until complete discharge ($\text{SOC} = 0$) is obtained by solving:

$$t_{\text{empty}} = \int_{\text{SOC}_0}^0 \frac{C_{\text{eff}}}{I_{\text{total}}(\text{SOC}, t)} d\text{SOC} \quad (5)$$

In practice, this is computed numerically due to the time-dependent nature of $I_{\text{total}}(t)$.

This foundational model enables quantitative predictions of battery life under various usage scenarios. In the following chapter, we apply it to specific cases, analyze the results, and conduct comprehensive model validation.

3.1.2 Factor Selection and Simplification Strategy

Thanks to [2], we're able to consider all kinds of

3.1.3 Environmental and Aging Factors

Lithium-ion battery capacity degrades with usage due to irreversible electrochemical changes during charge-discharge cycling. The aging correction factor f_{aging} depends on the accumulated cycle count N_{cycles} , where one cycle represents 100% cumulative depth-of-discharge.

3.2 Hardware Components Modeling

3.2.1 Display Power Model

While Carroll & Heiser's work [1] established a linear relationship between display power and backlight intensity for LCDs, this model is inapplicable to modern AMOLED displays. The fundamental difference lies in AMOLED's pixel-independent emission, which replaces uniform backlighting. We therefore adopt a more sophisticated model tailored to AMOLED's characteristics [3], whose formulation accurately captures the power consumption patterns of contemporary screens.

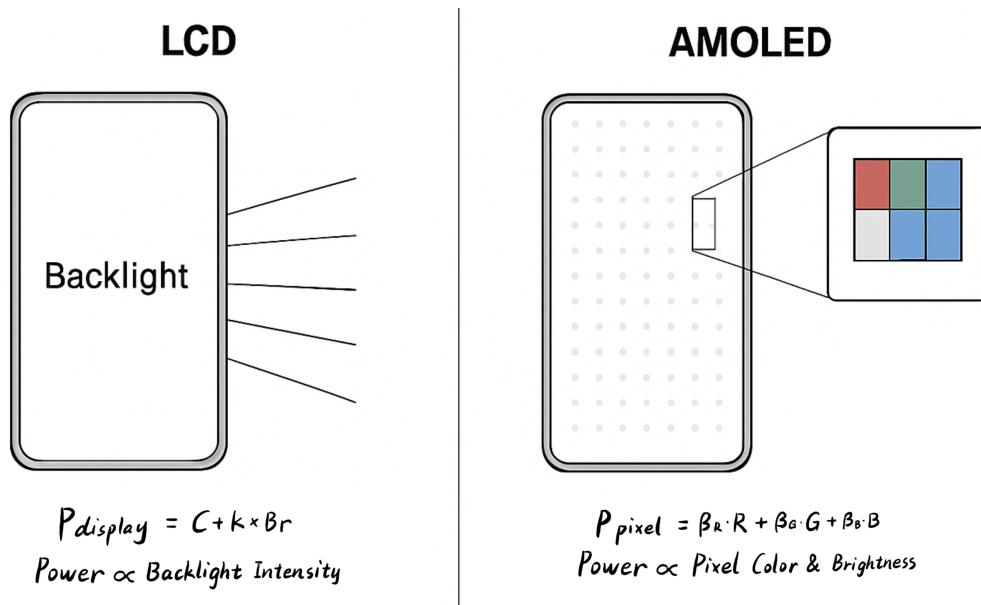


Figure 1: Schematic comparison of LCD and AMOLED power models.

The power consumption of a single AMOLED pixel is modeled as a function of its color intensity and the global brightness setting. The total display current I_{disp} is calculated by summing the contribution of all pixels and converting the total power to current:

$$I_{\text{disp}} = \frac{1}{V} [C + Br \cdot N \cdot (\beta_R R + \beta_G G + \beta_B B + a(R + G + B) + b)] \quad (6)$$

where $R, G, B \in [0, 1]$ are the normalized intensities of the red, green, and blue sub-pixels derived from $R_{\text{avg}}, G_{\text{avg}}, B_{\text{avg}}$. All other parameters are defined in Table 1.

Table 1: Parameters of the AMOLED Display Power Model

Symbol	Unit	Meaning
C	mW	Base power (black screen).
$\beta_R, \beta_G, \beta_B$	mW	Power coeff. per sub-pixel intensity.
a	mW	Linear coeff. for RGB sum correction.
b	mW	Constant for RGB sum correction.
$R_{\text{avg}}, G_{\text{avg}}, B_{\text{avg}}$	– (0-255)	Assumed average screen color.
N	–	Pixel count ($10^6, \approx 720\text{p}$).
V	V	System voltage.
Br	– (0.0–1.0)	Global brightness factor.

Equation (6) decomposes the display power into a fixed base cost C and a dynamic component scaled by the global brightness Br . The term inside the parentheses, $(\beta_R R + \beta_G G + \beta_B B + a(R + G + B) + b)$, represents the power consumed by a **single pixel** at full brightness. The coefficients $\beta_R, \beta_G, \beta_B$ capture the efficiency of each sub-pixel, while the linear correction term $a(R + G + B) + b$, identified in [3], is crucial for accurately modeling power at high luminance levels (e.g., white backgrounds common in applications). In contrast to the runtime screen analysis performed in the original study, **we simplify the model by assuming the screen displays a static, average color** ($R_{\text{avg}}, G_{\text{avg}}, B_{\text{avg}}$). This simplification allows for efficient system-level energy estimation without the need for real-time frame-buffer sampling, making it suitable for our integrated power model.

This model provides a physically-grounded method to estimate the AMOLED display's current draw as a function of user-defined brightness. Having established the display power model, we now turn to the processing units—the CPU and GPU—whose power consumption is governed by dynamic voltage and frequency scaling (DVFS).

- 3.2.2 Processor Power Model (CPU + GPU)**
- 3.2.3 Memory and Storage Model**
- 3.2.4 Sensor Power Model (GPS, Camera, etc.)**

3.3 Signaling Modules Modeling

- 3.3.1 Cellular and Wi-Fi Communication**
- 3.3.2 Bluetooth and Short-Range Communication**
- 3.3.3 Special Note: GPS as Positioning Sensor**

3.4 Software Role in the Modeling Framework

- 3.4.1 OS Overhead: Embedded in Hardware Parameters**
- 3.4.2 Application Impact: Hardware Usage Drivers**

4 Hardware-Software Parameter Calibration Methodology

5 Scenario-Based Analysis and Prediction

6 Model Analysis and Evaluation

[1]

References

- [1] Aaron Carroll and Gernot Heiser. “An analysis of power consumption in a smartphone”. In: *2010 USENIX Annual Technical Conference (USENIX ATC 10)*. 2010.
- [2] Kerry Hinton et al. “Power consumption and energy efficiency in the internet”. In: *IEEE Network* 25.2 (2011), pp. 6–12.
- [3] Russ Joseph and Margaret Martonosi. “Run-time power estimation in high performance microprocessors”. In: *Proceedings of the 2001 international symposium on Low power electronics and design*. 2001, pp. 135–140.

7 AI Use Report

Project Title: The River Intake Shield: Optimized Sensor Deployment and Emergency Response

Team Control Number: [2611750]