

# Real-time Programming Guidelines for C and C++

General, environment-agnostic principles (with examples and common pitfalls).

Generated on February 04, 2026.

This document summarizes key principles for writing real-time software. Real-time means correctness depends on both the produced value and meeting a deadline. The focus is determinism: predictable, bounded behavior, not just speed on average.

# 1) What "real-time" really means (and what it doesn't)

A real-time system is one where correctness depends on time as well as on producing the correct value.

- A "normal" program is correct if it computes the right output eventually.
- A real-time program is correct only if it computes the right output before a deadline.

## Hard vs soft real-time

- Hard real-time: missing the deadline is a system failure (for example, flight control, safety interlocks).
- Soft real-time: missing deadlines occasionally reduces quality (for example, video playback stutters) but is not catastrophic.

## Latency, jitter, and throughput

- Latency: the time from "event happens" to "your code responds".
- Jitter: the variation in latency or in periodic timing. If one cycle takes 1 ms and another takes 5 ms, jitter is high.
- Throughput: how much work per second. Throughput can be high even when jitter is terrible.

Real-time cares primarily about latency and jitter, especially the worst case, not the average.

## Determinism

Determinism means behavior is predictable: "This step always completes within 50 microseconds." Not: "Usually fast."

## Common misunderstanding (pitfall)

A system can have excellent average timing and still be non-real-time because it occasionally hiccups (allocator stalls, page faults, lock contention spikes, I/O stalls). Real-time engineering is largely about removing or bounding these hiccups.

# 2) Deadlines and WCET (Worst-Case Execution Time)

## WCET definition

WCET (Worst-Case Execution Time) is the maximum time a piece of code can take under worst conditions (worst inputs, worst cache state, worst contention, etc.).

Real-time developers try to:

- Design code so WCET is bounded and small.
- Measure and validate WCET under stress.

A key difference vs typical optimization: typical optimization aims to improve average runtime, while real-time optimization aims to reduce worst-case spikes and remove unpredictability.

## Time budgets (recommendation)

Think in time budgets. If your loop period is 1 ms, you might allocate:

- 100 microseconds for input acquisition
- 300 microseconds for compute
- 100 microseconds for output
- 200 microseconds for communications
- Leave margin (for example, 300 microseconds) for jitter, interrupts, cache misses, and rare paths

### **Rare-path spikes (pitfall)**

Worst case includes rare code paths like error handling, reconnection, buffer wrap-around, and first-time initialization behavior. Those often cause deadline misses in real systems.

## **3) Scheduling basics (how the system decides what runs next)**

In real-time, you typically have multiple tasks (threads or processes) competing for CPU time. Scheduling decides which one runs at each moment.

### **Common task types**

- Periodic tasks: run every fixed interval (for example, every 1 ms, every 20 ms). Example: a control loop that must execute every 5 ms.
- Event-driven tasks: run when something happens (for example, an input arrives, an interrupt triggers).
- Background or best-effort tasks: can run when CPU is free (logging, UI updates).

### **Priority**

Most real-time systems use priority-based scheduling: higher priority tasks preempt (interrupt) lower priority tasks.

The rule:

- Put "must meet deadlines" tasks at higher priority.
- Keep those tasks short and predictable.

### **Blocking in high-priority code (pitfall)**

A high-priority thread that sometimes blocks on a mutex (lock) is often worse than a slightly lower-priority thread that never blocks. Blocking destroys predictability.

### **Tiering (recommendation)**

Separate hard-deadline code from everything else. A simple model is:

- Tier 1: deadline-critical periodic loop (small, bounded)
- Tier 2: event processing (bounded per cycle)
- Tier 3: I/O, logging, UI, file or network output (best-effort)

## 4) Time-triggered vs event-triggered designs (two major styles)

### Time-triggered (cyclic executive) design

A cyclic executive is a fixed schedule like:

- Every 1 ms: read sensors
- Every 1 ms: compute control output
- Every 10 ms: send telemetry
- Every 100 ms: health check

Benefits: very predictable timing, easy to reason about worst case, great for hard real-time.

Downside: less responsive for sporadic events (unless you schedule frequent checks).

### Event-triggered design

Code runs in response to external events: a packet arrives, a digital input changes, an interrupt fires.

Benefits: very responsive; efficient when nothing happens.

Downside: harder to guarantee worst case under heavy event bursts unless you add strict bounds.

Most real systems use a hybrid: critical control is time-triggered; events are captured quickly and handled with bounded work.

### Bounded event handling (common pattern)

Even in an event-driven system, you usually add a rule like "process at most N events per loop iteration" so event storms cannot starve deadline-critical work.

## 5) Interrupts and ISRs (Interrupt Service Routines)

### Interrupt and ISR definition

An interrupt is a hardware signal that stops the CPU's current work and runs a special handler. The handler code is the ISR (Interrupt Service Routine).

ISR rules in real-time:

- Do as little as possible in the ISR.
- Capture minimal info (timestamp, a few bytes, set a flag).
- Defer heavier work to a normal task or thread.

Why: ISRs preempt normal code. A long ISR increases latency and jitter for everything else.

Typical pattern: ISR pushes data into a ring buffer (a fixed-size circular queue). A worker task drains the buffer and does processing.

### ISR work creep (pitfall)

Doing "just a little more work" in the ISR tends to grow over time (someone adds logging, then parsing, then validation). ISRs become accidental "God functions" and timing collapses.

## Conceptual example

ISR: copy 16 bytes into a preallocated ring buffer and increment a write index. Worker thread: parse the message, update a state machine, compute outputs.

## 6) Priority inversion (a classic real-time failure mode)

### What is priority inversion?

Scenario:

- High-priority task H needs a lock (mutex) held by low-priority task L.
- But medium-priority task M runs and keeps preempting L.
- L cannot run to release the lock.
- H waits and misses its deadline.

This is priority inversion: the low-priority task effectively blocks the high-priority one.

### How to mitigate

- Keep lock-holding times tiny.
- Use mutexes with priority inheritance: if L blocks H, L temporarily inherits H's priority so it can run and release the lock quickly.
- Prefer designs that minimize shared locks altogether.

### Rare still happens (pitfall)

Priority inversion can happen even if you rarely lock, because rare is still possible at exactly the wrong time.

### Single-writer ownership (pattern)

A common mitigation is single-writer ownership (one thread owns state; others send messages). That reduces locking and thus reduces priority inversion risk.

## 7) Memory management: the biggest source of unpredictability

### Why malloc/new are dangerous in real-time loops

Dynamic allocation (malloc, new) can:

- Take variable time (allocator searches free lists, coalesces blocks).
- Lock internally (contention).
- Fragment memory over long runtimes (slowdowns appear after hours or days).

Real-time rule: no dynamic allocation in real-time paths.

## Practical alternatives

- Preallocate everything at startup (buffers, objects, arrays).
- Fixed-size pools: allocate a pool of N objects, then take/return objects in constant time.
- Static storage: global or static arrays for fixed budgets.

## Paging and page faults

On systems with virtual memory, if memory is not currently in RAM, accessing it can trigger a page fault, which may take milliseconds.

Mitigation: touch memory during initialization so it is loaded; lock memory if supported; avoid lazy initialization in real-time threads.

## C++ hidden allocation pitfalls

Very common C++ pitfalls:

- `std::string` may allocate (even if it sometimes does not due to small string optimization).
- `std::vector` reallocates unless you reserve or fix capacity.
- `std::function` may allocate depending on what you store in it.
- Logging frameworks often allocate and/or take locks.
- First call to something may allocate (lazy init, internal caches, locale setup).

Practical rule: treat anything that might allocate as unsafe in deadline-critical code unless you can prove otherwise and lock it down.

# 8) Data structures and algorithms: bounded beats clever

Real-time requires predictable upper bounds.

## Avoid these in real-time paths

- Containers that sometimes resize or rehash: `std::vector` without reserved capacity; `std::unordered_map` (rehash spikes).
- Anything with potentially unbounded runtime: regex parsing; recursion with uncertain depth; "while queue not empty process all" when queue can grow without limit.

## Prefer these

- Fixed-size arrays or fixed-capacity vectors.
- Ring buffers.
- Pre-sized `std::vector` with reserve or built fixed capacity.
- Deterministic state machines.

## Bounded per cycle processing (key technique)

- In each cycle, process at most N items.
- If more arrive, defer to next cycle or drop (depending on requirement).

This prevents event storms from exploding execution time.

## Examples of boundedness

- Parse at most N packets per cycle.
- Run at most N iterations of a solver per cycle (and carry state forward).
- Coalesce events: if 500 position-update messages arrive, keep only the latest for this cycle.

## Big-O is not enough (pitfall)

Even bounded big-O can still be too large if constants are high or memory access is random (cache-unfriendly).

# 9) Concurrency: avoid shared mutable state when possible

## Why shared state is risky

Shared state requires synchronization: locks (mutexes) or atomic operations.

Locks can block unpredictably and cause priority inversion. Atomics avoid blocking but can be tricky and still cause contention.

## Safer architecture: ownership and message passing

- One thread owns a piece of data and is the only writer.
- Other threads communicate via messages (queues).
- Readers get snapshots (copies) or use double buffers.

This makes timing more predictable and simplifies reasoning.

## Common patterns

- SPSC ring buffer (Single Producer, Single Consumer): one writer thread, one reader thread; efficient and predictable.
- Double-buffering: one buffer is active for readers while the writer fills the other; then swap pointers or indices.
- Snapshot model: periodically copy a small state struct that readers use without locking.

## Tiny locks grow (pitfall)

A "tiny lock" becomes expensive if taken frequently, or if contention occurs at the wrong priority boundary.

## 10) Logging and I/O: keep it out of real-time threads

I/O (disk, network, console) can block unpredictably. Even formatting strings can allocate memory or take variable time.

Best practice: RT thread writes minimal binary records into a ring buffer; a background thread drains it and writes logs/telemetry.

### Recommendation

If you must log from real-time code, log only counters, fixed-size numeric fields, and timestamps, and avoid dynamic formatting.

### Async still bursts (pitfall)

Non-blocking sockets and async logging can still hide bursts: buffers fill and something eventually blocks or drops. That is acceptable if deliberate and bounded, not accidental.

## 11) Cache, branch prediction, and rare spikes

Even if your algorithm is bounded, hardware effects can produce jitter.

### Cache effects

If data or code is not in CPU cache, access is slower and worst-case latency can spike.

You reduce this by keeping hot code/data small and contiguous, and minimizing pointer-chasing linked structures.

### Branch prediction

CPUs try to guess which way a branch goes. Unpredictable branching can cause pipeline stalls.

In real-time, you care less about fastest and more about no surprises.

### Additional pitfalls and tips

- Avoid scattered allocations (linked lists, trees) in hot paths: pointer chasing causes cache misses.
- Watch for false sharing: two threads write different variables that happen to sit on the same CPU cache line; performance becomes jittery under load.
- Keep hot data structures aligned and compact.

## 12) State machines: the real-time developer's best friend

A state machine is a design where your system is in one of a finite number of states, and transitions happen only through defined events.

Benefits: predictable control flow, easier worst-case timing analysis, fewer unexpected behaviors. This is extremely common in embedded and real-time control.

### Example



- States: Idle, Armed, Running, Fault
- Events: StartCmd, StopCmd, SensorOk, SensorFail, Timeout
- Each event handler performs bounded work and updates state deterministically.

## Pitfall

Ad-hoc if spaghetti grows until you cannot reason about timing or correctness. State machines keep complexity controlled.

## 13) Testing and measurement: proving your timing

You cannot confidently claim real-time behavior without measurement.

Real-time teams track:

- Maximum observed cycle time
- Jitter
- Deadline misses
- Queue depths and overflow counts

Then they stress the system: maximum input rates, worst-case scenarios, long-duration runs (to catch fragmentation, drift, leaks).

## Recommendations

- Instrument start/end timestamps around every major stage in your loop.
- Track max observed per stage (not just average).
- Keep overrun counters and dropped event counters and treat them as first-class health metrics.

## Measurement distortion (pitfall)

If your measurement/logging is heavy, it can distort timing. Common pattern: collect lightweight counters in RT code and export them periodically from a lower-priority context.

## 14) Practical C/C++ rules of thumb for real-time code

### Coding rules

- No dynamic allocation in real-time loops.
- No blocking I/O in real-time threads.
- No unbounded loops ("process everything until empty") in RT paths.
- Prefer fixed-size buffers and bounded queues.
- Keep ISR minimal; defer work to threads.
- Keep critical sections tiny; avoid nested locks.
- Prefer explicit error handling over exceptions in RT paths.

## Optimization goals

- Reduce worst-case runtime.
- Reduce variance (jitter).
- Keep memory access predictable.
- Avoid hidden work (allocations, resizing, flushing, system calls).

## C/C++ pitfalls

- Hidden copies: passing large structs by value; unintended `std::string` copies.
- Hidden temporaries: operator overloading and expression templates can surprise you.
- Convenient abstractions that allocate or lock under the hood.
- Debug-only checks that are fine until someone compiles with them enabled in production.
- Floating-point corner cases: denormals/subnormals (extremely tiny numbers) can drastically slow some computations on some hardware; many systems flush them to zero to avoid timing spikes.

## 15) A bigger point: real-time is a system property

Even perfect C++ can fail deadlines if the environment introduces spikes (CPU frequency scaling, background services, scheduler behavior, driver latencies, interrupt storms).

So real-time engineering is always: architecture + code + OS configuration + measurement.

## Recommendation

Even if you are environment-agnostic, it helps to design assuming: preemption can occur at any time; interrupts can burst; memory access can be slower than expected; and some subsystem can stall unpredictably. Structure your code so stalls happen only in noncritical layers.

## 16) A concrete mental model (example)

Imagine a 1 kHz control loop (every 1 millisecond):

- Read inputs (must finish within 100 microseconds).
- Compute control (must finish within 300 microseconds).
- Write outputs (must finish within 100 microseconds).
- Housekeeping (optional, bounded, maybe 200 microseconds).

Total worst-case must remain below 1 millisecond with margin. Anything unbounded (allocations, logging flush, processing "all queued messages") risks blowing the budget.

## What margin is for

- Cache misses
- Interrupts
- Occasional slow branches

- Synchronization overhead
- Rare but valid states (fault paths, reconnection logic)

A good real-time design keeps margin intentionally, not accidentally.

## Added: Concrete mini-examples and patterns (C/C++ oriented)

### A) Ring buffer pattern (fixed capacity, predictable)

A ring buffer is a fixed-size queue where indices wrap around. Typical use: a producer (ISR or I/O thread) pushes fixed-size messages; a consumer (worker thread) pops and processes.

Why it is real-time friendly: no allocation; bounded operations (constant-time); explicit overflow behavior (TryPush returns false).

Pitfall: multi-producer/multi-consumer queues are much harder to implement correctly and predictably. If you need them, keep the design simple (for example, one producer per queue and merge later).

### Conceptual C++ sketch (Single Producer + Single Consumer) with abundant inline comments

```
#include <array>
#include <atomic>
#include <cstdint>

/*
    RingBufferSpSC<T, Capacity>

    SPSC = Single Producer, Single Consumer.

    Intent:
    - Provide a fixed-capacity (no allocations) queue.
    - Exactly one producer thread calls TryPush().
    - Exactly one consumer thread calls TryPop().

    Why this is real-time friendly:
    - No malloc/new during operation.
    - Each push/pop is constant-time (bounded).
    - Overflow/underflow is explicit via return value (no blocking).
*/

template<typename T, std::size_t Capacity>
class RingBufferSpSC
{
public:
    // TryPush:
    // - Returns false if the buffer is full (caller decides what to do: drop, count,
    // backpressure, etc.)
    // - Returns true if the item was successfully queued.
    bool TryPush(const T& item)
    {
        // Read the current write index (where the producer will write next).
        // memory_order_relaxed is fine for *this* load because the producer is the only
        // writer of m_write.
```

```

const std::size_t write = m_write.load(std::memory_order_relaxed);

// Compute the next write index (wrap around at Capacity).
const std::size_t next = Increment(write);

// Full condition:
// If advancing write would catch up to read, the ring is full.
//
// We use memory_order_acquire on the read index to ensure we observe the
// consumer's
// latest read position before deciding the buffer is full.
if (next == m_read.load(std::memory_order_acquire))
{
    return false; // Buffer full: caller must handle overflow policy.
}

// Store the item into the slot pointed to by "write".
// This is a plain assignment into preallocated storage (no allocation).
m_data[write] = item;

// Publish the new write index.
// memory_order_release ensures that the data write to m_data[] becomes visible to
// the consumer
// before the consumer observes the updated m_write.
m_write.store(next, std::memory_order_release);

return true;
}

// TryPop:
// - Returns false if the buffer is empty.
// - Returns true and writes the popped item to "out" if available.
bool TryPop(T& out)
{
    // Read the current read index (where the consumer will read next).
    // memory_order_relaxed is fine for *this* load because the consumer is the only
    // writer of m_read.
    const std::size_t read = m_read.load(std::memory_order_relaxed);

    // Empty condition:
    // If read equals write, there is nothing to pop.
    //
    // memory_order_acquire pairs with the producer's memory_order_release store to
    // m_write,
    // ensuring that if we observe an updated m_write, we also observe the
    // corresponding m_data[] write.
    if (read == m_write.load(std::memory_order_acquire))
    {
        return false; // Buffer empty.
    }

    // Copy the item out of the ring slot.
    // This should be a bounded operation; keep T small or trivially copyable if you
    // want tight timing.
    out = m_data[read];

    // Advance the read index (consumes this slot).
    // memory_order_release publishes the updated read index to the producer.
    m_read.store(Increment(read), std::memory_order_release);
}

```

```

        return true;
    }

private:
    // Increment:
    // - Wraps an index around Capacity.
    // - constexpr makes this foldable by the compiler for fixed Capacity.
    static constexpr std::size_t Increment(std::size_t idx)
    {
        return (idx + 1) % Capacity;
    }

    // Preallocated storage:
    // - Fixed size at compile time.
    // - No heap usage; avoids fragmentation and allocator latency.
    std::array<T, Capacity> m_data{};

    // Producer-owned index:
    // - Only the producer thread writes m_write.
    // - Consumer reads it to test empty / pop.
    std::atomic<std::size_t> m_write{0};

    // Consumer-owned index:
    // - Only the consumer thread writes m_read.
    // - Producer reads it to test full / push.
    std::atomic<std::size_t> m_read{0};
};

```

## B) Fixed-size object pool (avoid malloc/new in RT path)

Pattern: allocate N objects at startup; acquire/release from a free list. Benefits: bounded allocation time; no fragmentation.

Pitfall: watch for concurrency. You may need per-thread pools or lock-free free lists; otherwise you reintroduce lock contention.

## C) Double-buffering (predictable reads without locking)

Pattern: writer fills back buffer, then swaps a pointer/index so readers see the new front buffer. Readers always read a stable snapshot.

Good for sharing a current state struct (sensor snapshot, control parameters).

Pitfall: if state is large, copying every cycle may be too expensive; then use smaller snapshots or partial updates.

## D) Bounded per cycle event processing

Pseudo-logic: in each call, process at most N events; exit early if queue is empty. This guarantees an upper bound on the work performed per cycle.

## Conceptual C++ example with abundant inline comments

```

// Bounded per-cycle event processing
//
// Intent:
// - Prevent an "event storm" from consuming unbounded time in one cycle.

```

```

// - Guarantee an upper bound on work per cycle (max_events_per_cycle).
//
// Typical usage:
// - Run this once per periodic tick (e.g., every 1 ms, 10 ms, etc.)
// - Or run in an event loop that must remain responsive/predictable.

template<typename QueueT, typename EventT>
void ProcessEventsBounded(QueueT& queue)
{
    // Upper bound on events processed in this call.
    // Choose this based on your time budget and the worst-case cost of HandleEvent().
    constexpr int max_events_per_cycle = 50;

    for (int i = 0; i < max_events_per_cycle; ++i)
    {
        // Create a local event object. Keep it fixed-size to avoid allocations.
        EventT ev{};

        // Try to pop one event.
        // If the queue is empty, exit early.
        if (!queue.TryPop(ev))
        {
            break;
        }

        // HandleEvent must be designed to be bounded:
        // - no allocations
        // - no blocking I/O
        // - no unbounded loops
        //
        // If processing can be expensive, split it into smaller steps and carry state
        // across cycles.
        HandleEvent(ev);
    }

    // If events arrive faster than you process them, the queue may fill.
    // That is not "mysterious": you must define a policy (drop oldest, drop newest,
    // coalesce, etc.)
    // and track counters (drops, max queue depth, overruns) for observability.
}

```

Pitfall: if events arrive faster than you handle them, the queue fills. That is not a bug. You must define an overflow policy (drop newest, drop oldest, coalesce, compress) and track it with counters/telemetry.

## Added: Common failure modes (pitfalls) checklist

These are classic real-time problems that show up in code reviews:

- Unbounded loops: "process all pending messages" without a cap.
- Hidden allocations: string formatting, logging, containers resizing.
- Blocking calls: I/O, sleeps, waiting on condition variables in high priority code.
- Lock contention and priority inversion: high-priority thread waits on a mutex held by low-priority thread.
- Work creeping into ISRs: parsing and logging inside interrupt handlers.
- Rare-path spikes: reconnect logic, error handling, first-use lazy init.

- Poor cache behavior: pointer chasing, huge working set, false sharing.
- No backpressure plan: queues fill, latency explodes, or system thrashes.
- No measurement: "seems fine" until it fails under load.

## **Added: Real-time-friendly design habits that pay off**

- Make overflow explicit (return false, increment a drop counter).
- Structure code as stages with budgets (input, compute, output, optional).
- Use state machines for mode logic.
- Keep data local to a thread and communicate via messages.
- Pre-initialize everything (memory, tables, caches) before entering RT mode.
- Prefer simple, predictable code to fancy abstractions in RT paths.
- Treat "rare" as "will happen at the worst moment."