

# OPTIMUS v2.1: Performance-Optimized Direct API Integration with Progressive Loading for Large-Scale Data Selection in SolidJS Applications

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## Abstract

We present OPTIMUS v2.1, a performance-optimized approach to large-scale data selection in SolidJS applications that resolves critical browser freezing and CSS rendering bottlenecks. This paper documents the discovery and resolution of severe performance degradation caused by expensive CSS properties applied to hundreds of DOM elements simultaneously. Through systematic analysis of browser rendering behavior, we identified that `backdrop-filter`, `will-change`, and `contain` properties create catastrophic performance degradation when applied to large element sets. Our solution implements staggered style application using `requestAnimationFrame` scheduling and eliminates expensive CSS operations in favor of lightweight alternatives. OPTIMUS v2.1 achieves a 93% reduction in selection time (1527ms to 100ms for 216 items) and eliminates browser freezing entirely, while maintaining visual consistency and user experience quality.

## 1 Introduction

The original OPTIMUS system introduced optimistic UI updates to address the "select all" problem in large datasets. However, real-world deployment revealed fundamental limitations: inconsistent state between optimistic updates and actual data, network overhead from multiple API calls, and complexity in polling mechanisms. OPTIMUS v2 addressed these architectural issues with direct API integration and progressive loading, achieving excellent algorithmic performance.

However, production deployment of OPTIMUS v2 revealed an unexpected and severe performance crisis. Users reported complete browser freezing during selection operations on datasets as small as 200 items, with selection times exceeding 1.5 seconds for moderate datasets. The performance degradation

was so severe that system CPU fans would activate during selection operations, indicating excessive computational load.

Investigation revealed that the performance bottleneck was not algorithmic but rendering-related. Expensive CSS properties, particularly **backdrop-filter**, **will-change**, and **contain**, when applied to hundreds of DOM elements simultaneously, overwhelmed browser rendering pipelines. This discovery led to OPTIMUS v2.1, which focuses on CSS performance optimization through elimination of expensive properties and implementation of staggered style application.

OPTIMUS v2.1 achieves a 93.4% reduction in selection completion time (1527ms to 100ms), complete elimination of browser freezing, and maintains 60fps scroll performance with active selections. The optimization demonstrates that CSS rendering can be the primary performance bottleneck even when algorithmic efficiency is optimal.

## 2 System Architecture Evolution

### 2.1 From Optimistic to Direct: API Design

The core architectural shift in OPTIMUS v2 is the replacement of optimistic UI updates with direct API integration. The new `/api/browse-all` endpoint provides a lightweight, optimized response containing only essential metadata:

Listing 1: Browse-All API Response Structure

```
{
  "total_folders": 42,
  "total_images": 1337,
  "mtime": "2024-01-15T10:30:00Z",
  "items": [
    {"name": "folder1", "type": "directory"},
    {"name": "image1.jpg", "type": "image"},
    ...
  ]
}
```

This endpoint is specifically optimized for selection operations, returning only **name** and **type** fields without expensive image processing operations like thumbnail generation or metadata extraction.

### 2.2 Progressive Loading Architecture

The progressive loading system (`useProgressiveProcessor`) implements time-slicing to handle large datasets without blocking the main thread. The system uses a configurable time budget (default: 16ms) to maintain 60fps performance:

Listing 2: Progressive Processor Configuration

```
interface ProgressiveLoadingOptions {
  batchSize?: number;    // Items per batch (default: 1000)
```

```

    maxBatchTime?: number; // Max time per batch (default: 16ms)
    onProgress?: (current: number, total: number) => void;
}

```

The processor uses `requestAnimationFrame` for proper browser scheduling, ensuring smooth UI updates even during intensive processing operations.

## 2.3 Name-Based Selection System

OPTIMUS v2 implements a name-based selection system that provides consistency across pagination and eliminates the index-based conflicts present in the original design. The selection state maintains `Set<string>` collections for both images and folders:

Listing 3: Selection State Structure

```

interface SelectionState {
    multiSelected: Set<string>; // Selected image names
    multiFolderSelected: Set<string>; // Selected folder names
    selected: number | null; // Current focus index
    mode: "view" | "edit"; // Interaction mode
}

```

This approach ensures that selections persist correctly across page changes and eliminates the need for complex index-to-name mappings.

# 3 Algorithmic Implementation

## 3.1 Dual-Path Selection Algorithm

OPTIMUS v2 implements a dual-path algorithm that automatically selects the optimal processing strategy based on dataset size:

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**Algorithm 1** OPTIMUS v2 Select All Algorithm

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```
1: function SELECTALL
2:   Input: backendData, authFetch, app.notify, setState
3:   data  $\leftarrow$  backendData()
4:   if data is null then
5:     app.notify("No items to select", "info")
6:     return
7:   end if
8:   response  $\leftarrow$  await authFetch("/api/browse-all")  $\triangleright$  Direct API Call
9:   allItemsData  $\leftarrow$  await response.json()
10:  if allItemsData.items.length = 0 then
11:    app.notify("No items to select", "info")
12:    return
13:  end if
14:  totalItems  $\leftarrow$  allItemsData.items.length
15:  THRESHOLD  $\leftarrow$  5000
16:  if totalItems  $\leq$  THRESHOLD then  $\triangleright$  Fast Path: Direct Processing
17:    processDirectly(allItemsData.items)
18:  else  $\triangleright$  Progressive Path: Time-Sliced Processing
19:    processProgressively(allItemsData.items)
20:  end if
21:  app.notify("Selection complete", "success")
22: end function
23: function PROCESSDIRECTLY(items)
24:  batch(()  $\Rightarrow$  {
25:    newSelected  $\leftarrow$  new Set<string>()
26:    newFolderSelected  $\leftarrow$  new Set<string>()
27:    for item in items do
28:      if item.type = "image" then
29:        newSelected.add(item.name)
30:      else if item.type = "directory" and item.name  $\neq$  "." then
31:        newFolderSelected.add(item.name)
32:      end if
33:    end for
34:    setState({multiSelected: newSelected,
35:      multiFolderSelected: newFolderSelected})
36:  })
37: end function
38: function PROCESSPROGRESSIVELY(items)
39:  processor  $\leftarrow$  useProgressiveProcessor(items, itemProcessor, {
40:    batchSize: 1000,
41:    maxBatchTime: 16,
42:    onProgress: updateNotification
43:  })
44:  await processor.start()
45:  setState(processedResults)
46: end function
```

---

### 3.2 Time-Sliced Processing Implementation

The progressive processor implements sophisticated time management to maintain UI responsiveness:

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**Algorithm 2** Time-Sliced Processing

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```
1: function PROCESSBATCH
2:   while currentIndex  $\leq$  items.length and not cancelled do
3:     startTime  $\leftarrow$  performance.now()
4:     batchEndIndex  $\leftarrow$ 
5:       min(currentIndex + batchSize, items.length)
6:     while currentIndex  $\leq$  batchEndIndex and not cancelled do
7:       if performance.now() - startTime  $\geq$  maxBatchTime then
8:         break  $\triangleright$  Yield control to browser
9:       end if
10:      processItem(items[currentIndex])
11:      currentIndex  $\leftarrow$  currentIndex + 1
12:    end while
13:    updateProgress(currentIndex / items.length)
14:    if currentIndex  $\leq$  items.length then
15:      await requestAnimationFrame()  $\triangleright$  Yield to browser
16:    end if
17:  end while
18: end function
```

---

## 4 CSS Performance Crisis and Resolution

### 4.1 Performance Bottleneck Discovery

Real-world deployment of OPTIMUS v2 revealed an unexpected and severe performance crisis that manifested as complete browser freezing during selection operations. User reports indicated that selecting even moderate datasets (200+ items) resulted in:

- Complete UI freezing for 1-2 seconds during selection
- Extreme CPU usage causing system fans to activate
- Sluggish scrolling performance when large selections were active
- Browser responsiveness degradation affecting the entire tab

Performance monitoring revealed that a 216-item selection operation required 1527ms to complete - a performance rate of only 141 items/second, far below acceptable thresholds for responsive web applications.

## 4.2 Root Cause Analysis

Systematic investigation revealed that the performance degradation was not caused by JavaScript execution or API calls, but by CSS rendering operations. The original implementation applied expensive CSS properties to hundreds of DOM elements simultaneously:

Listing 4: Problematic CSS Implementation

```
.item.multi-selected {
  backdrop-filter: blur(2px) saturate(220%);
  will-change: transform, opacity;
  contain: layout style paint;
  position: relative;
}

.item.multi-selected::before {
  backdrop-filter: blur(2px) saturate(220%);
  /* Additional expensive pseudo-element styling */
}
```

The `backdrop-filter` property, while visually appealing, forces the browser to:

1. Create separate rendering contexts for each element
2. Apply expensive blur and saturation operations
3. Recalculate compositing layers for the entire viewport
4. Maintain these operations during scroll events

When applied to hundreds of elements simultaneously, these operations overwhelm the browser’s rendering pipeline, creating a cascade of performance degradation.

## 4.3 CSS Optimization Strategy

The solution involved a complete redesign of the selection styling approach:

Listing 5: Optimized CSS Implementation

```
/* Ultra-lightweight multi-selected styling */
.item.multi-selected {
  outline: 3px solid var(--accent);
  outline-offset: -3px;
  /* Removed all expensive properties */
}

.item.directory.multi-selected {
  background: color-mix(in srgb, var(--accent) 15%, var(--card-bg));
}
```

Key optimizations included:

- **Elimination of backdrop-filter:** Replaced with lightweight outline styling
- **Removal of pseudo-elements:** Eliminated expensive `::before` overlays
- **GPU hint removal:** Removed `will-change` and `contain` properties that become counterproductive at scale
- **Simplified color mixing:** Used efficient `color-mix()` for directory backgrounds

## 4.4 Staggered Style Application

For datasets where immediate styling could still cause performance issues, we implemented a staggered application system:

Listing 6: Deferred Style Application Algorithm

```
createEffect(() => {
  const isSelected = gallery.selection.multiSelected.has(props.idx);
  const totalSelected = gallery.selection.multiSelectedCount;

  if (isSelected && totalSelected > 50) {
    // Stagger application based on element index
    const delay = Math.floor(props.idx / 20) * 16; // 20 items per batch
    timeoutId = window.setTimeout(() => {
      requestAnimationFrame(() => {
        setDeferredMultiSelected(true);
      });
    }, delay);
  } else {
    // Apply immediately for small selections
    setDeferredMultiSelected(isSelected);
  }
});
```

This approach processes selection styling in batches of 20 items with 16ms intervals, maintaining 60fps performance while preventing browser lockup.

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**Algorithm 3** Staggered CSS Application Algorithm

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```
1: function APPLYSELECTIONSTYLES(items, threshold)
2:   Input: items - array of DOM elements, threshold - batch size threshold
3:   if items.length  $\leq$  threshold then           ▷ Small selection: apply
    immediately
4:     for item in items do
5:       item.classList.add("multi-selected")
6:     end for
7:     return
8:   end if                                     ▷ Large selection: apply in staggered batches

9:   BATCH_SIZE  $\leftarrow$  20
10:  BATCH_DELAY  $\leftarrow$  16                      ▷ milliseconds
11:  for i  $\leftarrow$  0 to items.length - 1 do
12:    delay  $\leftarrow$  floor(i / BATCH_SIZE)  $\times$  BATCH_DELAY
13:    setTimeout(()  $\Rightarrow$  {
14:      requestAnimationFrame(()  $\Rightarrow$  {
15:        items[i].classList.add("multi-selected")
16:      })
17:    }, delay)
18:  end for
19: end function
```

---

## 5 Performance Analysis

### 5.1 Network Optimization

OPTIMUS v2 achieves significant network efficiency improvements over the original implementation:

Metric	OPTIMUS v1	OPTIMUS v2
API Calls	$n$ pages	1 call
Data Transfer	Full metadata	Names only
Network Latency	$n \times$ RTT	$1 \times$ RTT
Consistency	Eventual	Immediate

Table 1: Network Performance Comparison

For a directory with 10,000 items across 100 pages, OPTIMUS v1 required 100 API calls, while OPTIMUS v2 requires only 1, representing a 99% reduction in network overhead.



## 5.2 UI Responsiveness

The time-slicing mechanism ensures consistent frame rates during large operations:

- **Small datasets ( $\leq 5,000$  items):** Processed in single frame using `batch()`
- **Large datasets ( $> 5,000$  items):** Time-sliced with 16ms budget per frame
- **Progress feedback:** Real-time updates via notification system
- **Cancellation support:** Immediate response to user interruption

## 5.3 Memory Management

OPTIMUS v2 implements intelligent memory management for large selections:

Listing 7: Progressive localStorage Implementation

```
const LARGE_SELECTION_THRESHOLD = 1000;
if (totalSelections > LARGE_SELECTION_THRESHOLD) {
  const { saveProgressive } = useProgressiveLocalStorage();
  await saveProgressive("gallerySelection", stateToSave);
} else {
  localStorage.setItem("gallerySelection",
    JSON.stringify(stateToSave));
}
```

This approach prevents UI blocking during state persistence while maintaining data consistency.

# 6 Implementation Details

## 6.1 SolidJS Integration

OPTIMUS v2 leverages SolidJS's fine-grained reactivity through strategic use of `batch()` operations:

Listing 8: Batched State Updates

```
batch(() => {
  setState((prev) => ({
    ...prev,
    multiSelected: newMultiSelected,
    multiFolderSelected: newMultiFolderSelected,
  }));
  saveState();
});
```

This ensures that complex state updates trigger only a single reactivity cycle, minimizing rendering overhead.

## 6.2 Error Handling and Resilience

The system implements comprehensive error handling with graceful degradation:

Listing 9: Error Handling Strategy

```
try {
  const response = await authFetch(`/api/browse-all?path=${path}`);
  if (!response.ok) {
    throw new Error(`HTTP ${response.status}:
      ${response.statusText}`);
  }
  // Process response...
} catch (error) {
  console.error("Selection error:", error);
  app.notify("Failed to select all items", "error");
  // Maintain existing selection state
}
```

## 7 Real-World Performance Results

The CSS optimization efforts produced dramatic performance improvements measured in production environments:

### 7.1 Selection Performance Comparison

Metric	Original v2	v2.1 Optimized	Improvement	Target
216 items selection	1527ms	~100ms	93.4%	~100ms
Browser freeze time	1-2s	0ms	100%	0ms
Scroll FPS (selected)	15-20fps	60fps	300%	60fps
CPU usage spike	High	Normal	N/A	Normal
Memory efficiency	Poor	Excellent	N/A	Efficient

Table 2: CSS Performance Optimization Results

### 7.2 Scalability Analysis

The optimized approach maintains consistent performance across varying dataset sizes:

Dataset Size	Selection Time	Browser Freeze	Scroll Performance
200 items	~50ms	None	60fps
500 items	~100ms	None	60fps
1,000 items	~150ms	None	60fps
2,000 items	~250ms	None	58fps
5,000 items	~400ms	None	55fps

Table 3: Scalability Performance Analysis

### 7.3 User Experience Metrics

Post-optimization user feedback indicates complete resolution of the performance crisis:

- **Browser responsiveness:** No reported freezing incidents
- **System impact:** Elimination of CPU fan activation during selection
- **Scroll fluidity:** Smooth scrolling maintained even with large active selections
- **Selection speed:** Perceived as instantaneous for datasets under 1,000 items

The performance improvements demonstrate that CSS optimization can be more impactful than algorithmic improvements when dealing with browser rendering bottlenecks.

## 8 Future Enhancements

### 8.1 WebWorker Integration

Future versions may implement WebWorker-based processing for CPU-intensive operations:

Listing 10: WebWorker Processing Concept

```
const worker = new Worker('/workers/selection-processor.js');
worker.postMessage({ items: largeItemSet, operation: 'selectAll' });
worker.onmessage = (event) => {
  const { results, progress } = event.data;
  updateSelectionState(results);
};
```

## 8.2 Streaming API Integration

For extremely large datasets, streaming API integration could provide incremental updates:

Listing 11: Streaming Selection Concept

```
const stream = await fetch('/api/browse-all-stream');
const reader = stream.body.getReader();
while (true) {
  const { done, value } = await reader.read();
  if (done) break;
  processStreamChunk(value);
}
```

## 9 Conclusion

OPTIMUS v2.1 represents a critical breakthrough in understanding and resolving browser performance bottlenecks in large-scale data selection applications. The project's evolution from architectural optimization (v2) to CSS performance optimization (v2.1) demonstrates that rendering performance can be the limiting factor even when algorithmic efficiency is optimal.

### 9.1 Key Discoveries

The research revealed several important findings about browser performance at scale:

1. **CSS Property Scaling:** Properties like `backdrop-filter`, `will-change`, and `contain` become performance liabilities when applied to hundreds of elements simultaneously
2. **Rendering Pipeline Saturation:** Browser rendering pipelines have finite capacity that can be overwhelmed by seemingly reasonable operations
3. **Staggered Application Benefits:** Temporal distribution of style applications maintains 60fps performance during large operations
4. **User Perception vs. Technical Metrics:** Elimination of browser freezing has higher user impact than raw selection speed improvements

### 9.2 Performance Achievements

OPTIMUS v2.1 achieves:

- 93.4% reduction in selection completion time (1527ms to 100ms)
- Complete elimination of browser freezing incidents

- Maintenance of 60fps scroll performance with active selections
- Scalable performance across datasets up to 5,000+ items

### 9.3 Broader Implications

The CSS performance crisis and resolution in OPTIMUS provides valuable insights for the broader web development community:

- **Performance optimization priorities:** CSS rendering optimization can exceed algorithmic improvements in impact
- **Modern CSS caution:** Advanced CSS features require careful consideration when applied at scale
- **User-centric metrics:** Browser responsiveness and system impact are critical UX factors
- **Progressive enhancement:** Staggered application techniques can maintain performance while preserving visual quality

The OPTIMUS v2.1 implementation provides a robust foundation for building responsive data-intensive applications that maintain excellent performance characteristics even under extreme selection scenarios. The lessons learned about CSS performance scaling are applicable beyond selection operations to any web application dealing with large numbers of styled elements.

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