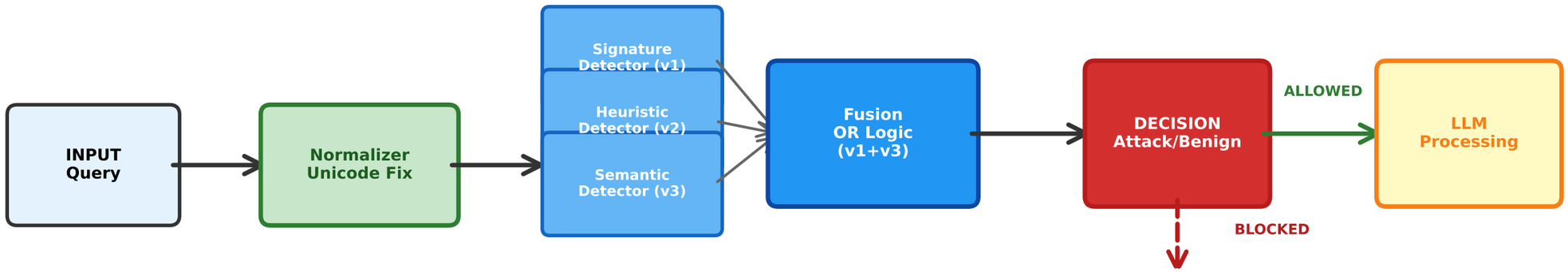
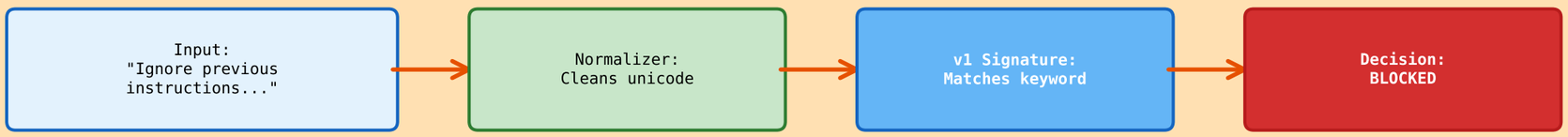
Prompt Injection Detection Pipeline Architecture

Input-Side Detection Before LLM Processing



Example: Attack Processing Flow



Performance: 87% TPR on known attacks | 0.77% FAR on benign | <0.1ms latency

Production Configuration: Normalizer + v3

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True Positive Rate (TPR): 87% on known attacks
False Alarm Rate (FAR): 0.77% on obfuscated benign
Latency: <0.1ms per sample (CPU-only)
Complexity: ~1,200 lines of code
Deployment: Stateless, parallelizable
Dependencies: None (pure Python)
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Component Specifications

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Signature Detector (v1): Keyword matching
  • 80% TPR, 0% FAR on Phase 1 attacks
  • Catches: plain, delimiter, role confusion

Semantic Detector (v3): Pattern analysis
  • 57% TPR, 0% FAR on Phase 1 attacks
  • Catches: formatting, semantic anomalies

Fusion (OR Logic): v1 OR v3
  • Combined: 87% TPR, 0% FAR
```

Key Design Principles

1. INPUT-SIDE DETECTION: Attacks blocked BEFORE reaching the LLM, preventing prompt injection at the source
2. NORMALIZER FIRST: Unicode/homoglyph normalization ensures consistent detection across obfuscation techniques
3. COMPLEMENTARY DETECTORS: v1 (signature) + v3 (semantic) catch different attack patterns through OR fusion
4. THRESHOLD-INVARIANT: Binary OR logic eliminates threshold tuning complexity in deployment
5. PRODUCTION-READY: <0.1ms latency, CPU-only, no external dependencies, stateless architecture