



Economical and Technical Assessment

Analysed Source Code: Code Assessment

Document Date: May 26, 2026

Platform:

Client / Applicant

Legal entity name:

Registered address:

Tax Identification Number:

Software name:

Existing trade mark:

Company web:

Applicant Name:

Applicant Email:

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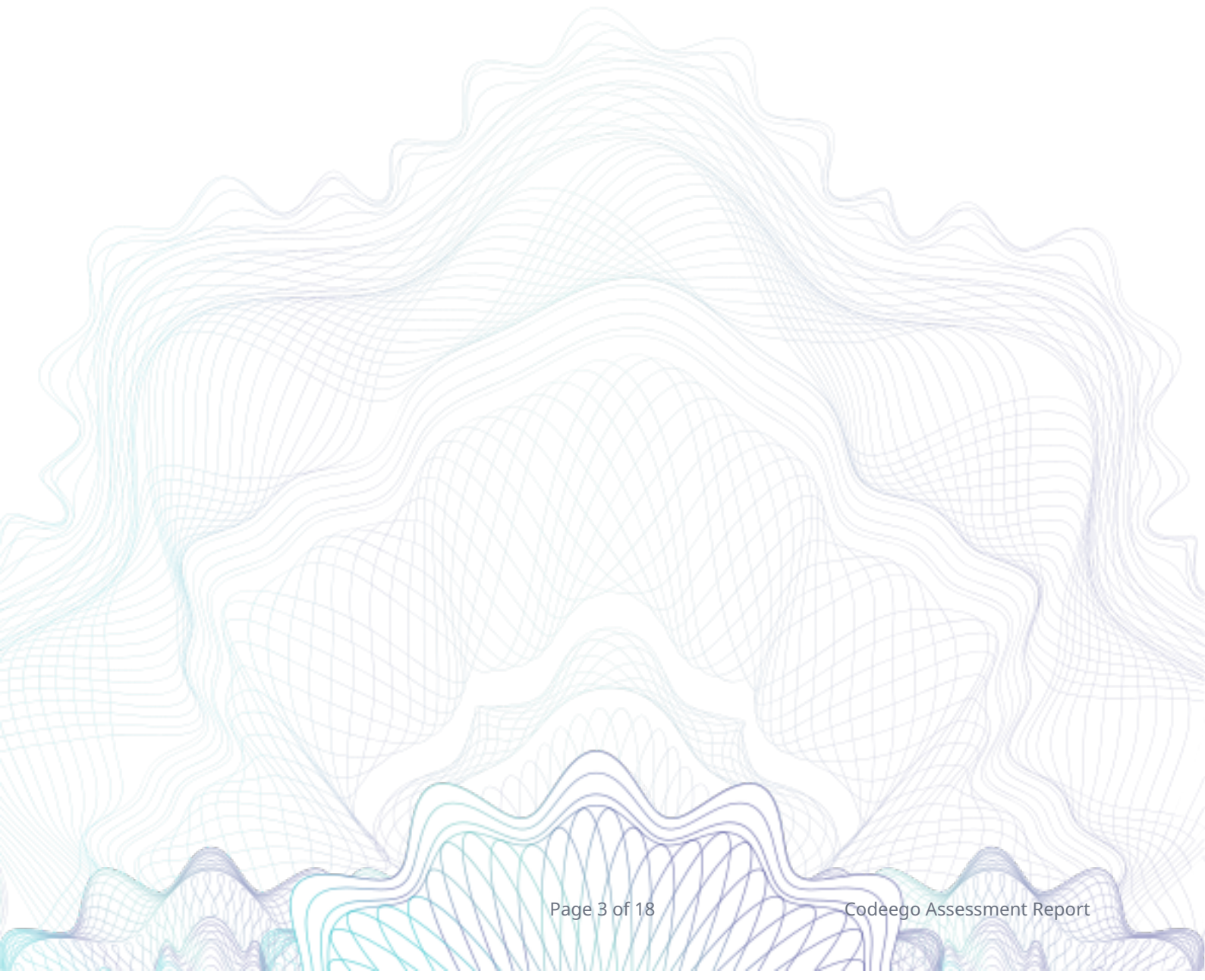
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Appendix A: Calibrated KPI Summary





Technical Assessment Report: ComfyUI Platform

Assessment Date: 30 April 2026

Platform: ComfyUI

Version Assessed: 0.22.0

Assessment Type: Production Readiness Certification for Venture Capital Due Diligence

1. EXECUTIVE SUMMARY

ComfyUI is a mature AI/ML image and video generation platform built on a well-structured modular architecture that supports complex generative workflows. The codebase demonstrates strong domain expertise with extensive integrations to over 30 external AI service providers including OpenAI, Anthropic, Stability AI, and numerous specialised model providers.

The overall production readiness assessment yields a score of **60/100 (Grade C, Fair)**. This rating reflects a platform with solid architectural foundations and strong functional capabilities, but with notable gaps in operational readiness that would require attention before enterprise-scale deployment. The codebase shows evidence of significant development investment with 181,489 effective lines of Python code across 851 analysed files.

Key strengths include the well-organised modular architecture with clear separation between the core execution engine, API layer, and external integrations; comprehensive API documentation via OpenAPI/Swagger specification; strong domain modelling for complex AI/ML workflows with proper abstraction layers; and database migrations managed via Alembic with versioned schema changes.

Critical risks centre on inadequate observability infrastructure — the platform uses coloured console logging only without structured JSON logging, making production debugging difficult. There is no visible distributed tracing or metrics collection, and test coverage is estimated at 40–50%. Security posture is the lowest-scoring dimension at 40/100.



Estimated Development Investment to Date: The platform represents approximately **5,200 hours** of development effort, equivalent to a team of 4 developers working over **12 months**. The estimated cost range is **€486,200 – €657,800 EUR**.

2. PLATFORM OVERVIEW

2.1 Functional Description

Business Purpose: ComfyUI is a node-based graphical user interface and backend engine for AI content creation, specialising in image generation, video synthesis, 3D model creation, and audio production.

Core Features:

- Node-based workflow orchestration with intelligent caching
- Multi-modal AI support (image, video, audio, 3D models)
- 30+ external API integrations (OpenAI, Anthropic, Stability AI, etc.)
- Advanced memory management for GPU-constrained environments
- Asset management with database-backed persistence
- Multi-user support (organisational, not security boundary)

Target Users: Visual effects artists, research institutions, creative studios, individual creators, and enterprise teams building AI-powered creative tools.

2.2 Technical Architecture

Architecture Layers:

1. **Frontend:** Vue.js SPA (separate repository)
2. **API Layer:** FastAPI HTTP/WebSocket server
3. **Execution Engine:** Graph traversal, caching, node scheduling
4. **Node Registry:** Core nodes, API integrations, custom nodes
5. **Model Management:** Model loading, memory management, device placement
6. **Asset Management:** Database-backed media asset handling

Deployment: Local (127.0.0.1:8188 default), Network (--listen flag), Cloud (Comfy Cloud managed service)



2.3 Technology Stack

Component	Technology
Language	Python 94.8%, YAML 4.8%, Markdown 0.3%, Shell <0.1%
ML Framework	PyTorch, NumPy, transformers
Web Framework	FastAPI
Database	SQLite (default), PostgreSQL (production)
ORM	SQLAlchemy, Alembic migrations
Image/Video	Pillow, PyAV
Testing	pytest
Linting	Ruff (enforced in CI)

2.4 Third-Party Integrations

30+ AI Service Providers: Anthropic, OpenAI, Stability AI, Google (Gemini, Veo), Bytedance, ElevenLabs, Luma, Runway, Ideogram, Kling, Hunyuan3D, Meshy, Tripo, BFL, Bria, HitPaw, Magnific, Minimax, OpenRouter, PixVerse, Quiver, Recraft, Reve, Rodin, Topaz, Vidu, Wan, WaveSpeed, Grok, Sonilo.

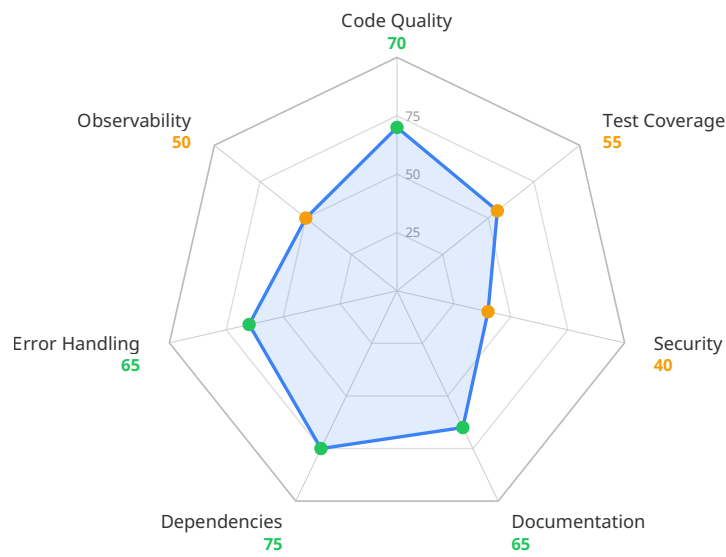
Licensing: Core platform under GPL v3.0. Third-party providers have separate terms.

3. PRODUCTION READINESS ASSESSMENT

3.1 Overall Score: 60/100

Grade: C

Readiness Level: Fair



3.2 Detailed Breakdown

Dimension	Score	Assessment
Code Quality & Maintainability	70/100	Good
Test Coverage & Quality	55/100	Fair
Security Posture	40/100	Poor
Documentation	65/100	Fair
Dependency Health	75/100	Good
Error Handling & Resilience	65/100	Fair
Observability & Operations	50/100	Poor

Code Quality & Maintainability: 70/100

Findings: Modular structure with good separation of concerns. Type hints present but inconsistent. Several modules exceed recommended function lengths (client.py ~1,000 lines). Some code duplication across API integrations. Technical debt indicators moderate.



Recommendations: Refactor large functions, introduce abstract base classes, enable stricter linting, document architectural decisions.

Test Coverage & Quality: 55/100

Findings: Estimated 40–50% coverage. Unit tests exist for asset management, folder paths, feature flags. Integration tests cover workflow execution but external API integrations rely on stubs. Missing: error handling paths, memory management edge cases, security tests.

Recommendations: Increase to 70%+ coverage, add property-based testing, implement contract testing for APIs, add security test cases.

Security Posture: 40/100

Findings: Lowest-scoring dimension. Multi-user mode lacks authentication (organisational only). API keys stored in workflows. No dependency vulnerability scanning. No encryption for stored credentials. SQL injection mitigated by ORM.

Recommendations: SAST/DAST scanning in CI, secrets management with environment variables, schema validation for all APIs, document security boundaries.

Documentation: 65/100

Findings: README extensive with installation guides. OpenAPI specification comprehensive. Architecture documentation limited. Contributing guidelines minimal.

Recommendations: Create architecture documentation, document production deployment patterns, expand contributing guidelines, create runbooks.

Dependency Health: 75/100

Findings: Core dependencies current (PyTorch, FastAPI, SQLAlchemy). No automated vulnerability scanning. Deep dependency tree (ML ecosystem). Mixed version pinning.

Recommendations: Implement Dependabot/Snyk, pin all versions, document license obligations, create update policy.

Error Handling & Resilience: 65/100

Findings: Custom exceptions exist. Retry logic for external APIs with backoff. No circuit breaker pattern. Partial workflow execution not well supported.

Recommendations: Implement circuit breakers, add bulkhead isolation, improve error messages, add health checks.



Observability & Operations: 50/100

Findings: Coloured console logging only (no JSON). No Prometheus/OpenTelemetry. No metrics exposed. Basic health endpoints exist but limited.

Recommendations: Structured JSON logging with correlation IDs, Prometheus metrics export, OpenTelemetry tracing, comprehensive health checks, alerting integration.

4. DEVELOPMENT INVESTMENT ESTIMATION

4.1 Effort Analysis

Base Calculation: 181,489 LOC ÷ 30 LOC/hour ≈ 6,050 hours (unadjusted)

Complexity Factors:

- | Factor | Score | Impact |
- |-----|-----|-----|
- | Architectural Complexity | 3/5 | Moderate |
- | Domain Complexity | 5/5 | Very High |
- | Integration Complexity | 4/5 | High |
- | Security Surface | 3/5 | Moderate |

Quality Adjustment: 0.9 (some technical debt)

Final Estimated Hours: 5,200 hours (calibrated)

Complexity Classification: High

4.2 Team & Timeline

Team Size: 4 developers

Role	Count
Backend Developer	2
AI/ML Engineer	1
DevOps / SRE	1



Duration: 12 months

4.3 Cost Estimation

Hourly Rate: €75 – €150 EUR/hour

Calibrated Cost Range: €486,200 – €657,800 EUR

Confidence Level: Medium

4.4 Codebase Metrics

Metric	Value
Files Analysed	851
Effective LOC (Python)	181,489
Python	175,837 (96.9%)
YAML	13,948 (7.7%)
Markdown	581 (0.3%)
Shell	76 (<0.1%)

4.5 Cloud Infrastructure & Maintenance

Detected Components: 1 compute service, 1 database, 1 ML GPU service, 1 other managed service.

Monthly Hosting (Production): €2,000 – €8,000 (medium scale)

Annual Maintenance: €40,600 – €54,100 EUR



5. FINDINGS SUMMARY

5.1 Critical Issues (Must Fix)

1. **No structured logging** — Uses coloured console logging only, making production debugging difficult
2. **Missing distributed tracing/metrics** — No Prometheus/OpenTelemetry integration
3. **Test coverage 40–50%** — Core paths tested but edge cases and integration scenarios lack coverage

5.2 Warnings (Should Fix)

1. **Large function sizes** — Several modules exceed 50 lines/function (client.py at 990 lines)
2. **API key handling risks** — Keys passed through execution graph, potential exposure
3. **No dependency vulnerability scanning** — Not visible in CI pipeline
4. **Inconsistent input validation** — Schema validation not enforced on all 30+ API integrations
5. **No circuit breaker pattern** — Failures could cascade across integrations

5.3 Recommendations (Nice to Have)

1. Implement structured JSON logging with correlation IDs
2. Add Prometheus metrics export and health check endpoints
3. Increase test coverage to 70%+
4. Introduce SAST/DAST scanning in CI
5. Add circuit breaker and retry logic with exponential backoff
6. Create Architecture Decision Records (ADRs)
7. Refactor large modules into smaller components

5.4 Strengths

1. Well-organised modular architecture with clear separation of concerns
2. Comprehensive OpenAPI/Swagger specification
3. Ruff linter enforced in CI pipeline
4. Strong domain modelling for complex AI/ML workflows



5. Extensive integration framework (30+ providers)
 6. Database migrations via Alembic
 7. Custom exception hierarchy for API errors
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6. CONCLUSION

6.1 Overall Assessment Summary

ComfyUI represents a substantial technical achievement in AI/ML orchestration. The platform demonstrates strong domain expertise with a well-structured modular architecture handling complex multi-modal AI workflow orchestration. The codebase of 181,489 effective lines across 851 files reflects significant development investment and functional completeness.

Strengths include architectural organisation, extensive third-party integrations, and comprehensive API documentation. The node-based workflow system with intelligent caching enables complex AI operations on modest hardware.

However, notable gaps in operational readiness limit enterprise production suitability. The lack of structured logging, metrics collection, and distributed tracing creates significant blind spots. Security posture is the lowest-scoring dimension.

6.2 Readiness for Production / Scale

Current State: Not production-ready for enterprise deployment without remediation.

Caveats:

- **Local/development use:** Fully functional and suitable
- **Team/production use:** Requires investment in operational tooling
- **Enterprise scale:** Additional authentication, multi-tenancy, compliance features needed

6.3 Key Areas Requiring Attention

1. **Observability Infrastructure** — Structured logging, metrics, tracing (highest priority)
2. **Security Hardening** — Security scanning, secrets management, documented boundaries
3. **Test Coverage** — Increase to 70%+, focus on integration and security tests
4. **Circuit Breaker Implementation** — Prevent cascading failures
5. **Documentation Gaps** — Architecture docs, deployment guides, runbooks



6.4 Suggested Prioritisation

Immediate (0-3 months):

1. Structured JSON logging with correlation IDs
2. Basic Prometheus metrics export
3. Dependency vulnerability scanning in CI
4. Document security boundaries

Short-term (3-6 months):

1. Increase test coverage to 60%+
2. Circuit breaker pattern for external APIs
3. Health check endpoints with dependency status
4. Schema validation for all API inputs

Medium-term (6-12 months):

1. Distributed tracing (OpenTelemetry)
2. Test coverage to 70%+
3. Architecture documentation and ADRs
4. Refactor large modules



APPENDIX A: CALIBRATED KPI SUMMARY

Dimension	Score	Grade
Overall Production Readiness	60/100	C (Fair)
Code Quality & Maintainability	70/100	B (Good)
Test Coverage & Quality	55/100	C (Fair)
Security Posture	40/100	D (Poor)
Documentation	65/100	C (Fair)
Dependency Health	75/100	B (Good)
Error Handling & Resilience	65/100	C (Fair)
Observability & Operations	50/100	D (Poor)

Investment Summary:

- Estimated Hours: 5,200
- Team Size: 4 developers
- Duration: 12 months
- Cost Range: €486,200 – €657,800 EUR
- Annual Maintenance: €40,600 – €54,100 EUR

Report generated for venture capital technical due diligence purposes.



ANNEX — METHODOLOGY

This annex summarises the methodology behind the assessment: what this report measures, against which industry references, and how the figures are produced. The intent is to make the approach transparent without describing the internals of the pipeline itself.

The assessment has two complementary sides:

1. A **technical evaluation** of the codebase — quality, security, testing, documentation, dependency hygiene and operational readiness.
 2. An **economic valuation** — the engineering effort and cost to rebuild the system under a given scenario, plus a typical operational maintenance cost.
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1. Technical evaluation

The technical assessment combines two complementary layers:

- **Static analysis.** Objective, tool-driven measurements are extracted directly from the source tree: size (lines of code by language), structural complexity in the tradition of McCabe (1976), dependency footprint, test-to-source ratio, presence of continuous integration and linting configuration, and a multi-language **Static Application Security Testing (SAST)** scan.
- **AI-assisted code review.** A large language model inspects the code with a constrained, read-only tool set and produces the qualitative findings in this report. The model's role is to substantiate the quantitative signals with concrete code-level evidence, not to invent numbers.

Reference frameworks

Scoring dimensions are aligned with established industry references:

Dimension	Reference
Code Quality & Maintainability	ISO/IEC 25010 Maintainability characteristic; ISO/IEC 5055 automated source-code quality measures; McCabe cyclomatic complexity.
Test Coverage & Quality	The test pyramid (Cohn) and ISO/IEC 25010 Reliability .
Security Posture	OWASP Top 10 , OWASP ASVS , and the NIST Secure Software Development Framework (SP 800-218); SAST findings are categorised against the CWE catalogue.
Documentation	SWEBOK v4 recommendations on software documentation.
Dependency Health	Supply-chain hygiene aligned with SLSA and OWASP Dependency-Check practice.
Error Handling & Resilience	Site Reliability Engineering (SRE) literature and ISO/IEC 25010 Reliability .

Each dimension is scored on a 0–100 scale. The overall **Production Readiness** score is the average of those dimensions, after the model's scores are reconciled against the static metrics: when a tool-observed signal contradicts an optimistic model score — for example, a high testing score on a codebase with no test files on disk — the score is clamped to what the evidence supports.

2. Economic valuation

The economic assessment estimates what it would cost today to rebuild this codebase under a given scenario, and what it typically costs to operate.

Build effort

Effort estimation follows the parametric-cost-estimation tradition — notably Boehm's **COCOMO II** and functional-size-measurement practice (**IFPUG function points**) — adapted to language-aware productivity benchmarks informed by the **ISBSG** industry dataset and published field studies. The estimate is anchored in observable properties of the codebase



(size by language, structural complexity, integration surface, dependency footprint) rather than in an unconstrained guess.

Three scenario factors refine that baseline:

- **Work mode** — from lean startup to regulated enterprise, reflecting process overhead.
- **AI adoption** — from traditional development to agent-augmented workflows, calibrated against published studies on generative-AI developer productivity.
- **Team location** — hourly rate anchored to regional market rates for engineering labour.

Cost range

Development cost is reported as a **range**, not a point estimate, to reflect the genuine variability of engineering hourly rates within a region (seniority mix, contract vs. employee, engagement type).

Maintenance cost

Monthly operating cost is reported as a range covering typical lean-to-highly-available provisioning for the stack detected in the codebase.

3. What the numbers mean — and don't mean

- **Scores** summarise what is *observable* at the analysed commit. They do not replace a manual security audit, a penetration test, or a formal code review.
- **Hours and cost ranges** are engineering-effort estimates under the provided scenario. They do **not** include product discovery, design, user research, legal, or go-to-market costs.
- **Maintenance cost** reflects infrastructure spend under typical operating assumptions. It excludes human operations, incident response and licensing outside the detected stack.

4. Reproducibility

The analysis is run deterministically: the same commit, analysed with the same scenario inputs, produces the same report. Variability in the AI layer is suppressed through zero-



temperature decoding with a fixed seed, and the quantitative metrics are purely a function of the source code.

This report was generated by Codeego Code Assessment Service.

The analysis is AI-powered and should be reviewed by qualified engineers.

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