

📁 Package Structure

This document explains the organization of the Constrained Intelligence Constants codebase.

Directory Tree

constrained-intelligence-constants/	
└── constrained_intelligence/	# Main package source code
├── __init__.py	# Package initialization and exports
├── core.py	# Core measurement and optimization classes
├── discovery.py	# Constant discovery algorithms
└── constants.py	# Fundamental constants definitions
└── tests/	# Test suite
├── __init__.py	# Tests for core module
├── test_core.py	# Tests for discovery module
├── test_discovery.py	# Tests for constants module
└── test_constants.py	
└── examples/	# Usage examples
├── basic_usage.py	# Simple examples for getting started
├── advanced_examples.py	# Complex use cases
└── notebook.ipynb	# Interactive Jupyter notebook
└── validation/	# Experimental validation
└── experimental_validation.py	# Validation experiments and benchmarks
└── docs/	# Documentation (Sphinx)
├── conf.py	# Sphinx configuration
├── index.rst	# Documentation index
└── api/	# API reference docs
└── .github/	# GitHub-specific files
└── workflows/	# GitHub Actions workflows
└── ci.yml	# Continuous integration pipeline
└── README.md	# Main project README
└── QUICKSTART.md	# Quick start guide
└── THEORY.md	# Mathematical theory and proofs
└── CONTRIBUTING.md	# Contribution guidelines
└── PACKAGE_STRUCTURE.md	# This file
└── CODE_OF_CONDUCT.md	# Code of conduct
└── CHANGELOG.md	# Version history
└── LICENSE	# MIT License
└── setup.py	# Package installation configuration
└── requirements.txt	# Runtime dependencies
└── requirements-dev.txt	# Development dependencies
└── MANIFEST.in	# Additional files for distribution
└── Dockerfile	# Docker container definition
└── .dockerignore	# Docker ignore patterns
└── .gitignore	# Git ignore patterns
└── .flake8	# Flake8 linter configuration

Core Modules

`constrained_intelligence/__init__.py`

Purpose: Package entry point, defines public API

Exports:

- Main classes: `ConstantsMeasurement`, `OptimizationEngine`, `BoundedSystemAnalyzer`
- Discovery tools: `ConstantDiscovery`, `DiscoveryMethods`
- All constants from `constants.py`

Usage:

```
from constrained_intelligence import ConstantsMeasurement, GOLDEN_RATIO
```

`constrained_intelligence/core.py`

Purpose: Core functionality for measurement, optimization, and analysis

Classes:

1. **MeasurementResult** (dataclass)
 - Stores results of constant measurements
 - Attributes: `constant_value`, `confidence`, `bounds`, `empirical_evidence`, `theoretical_basis`
2. **ConstantsMeasurement**
 - Measures constants in bounded systems
 - Methods:
 - `measure_resource_allocation()` : Find optimal resource split
 - `measure_learning_convergence()` : Predict convergence timing
 - `measure_optimization_efficiency()` : Analyze efficiency bounds
3. **OptimizationEngine**
 - Optimization algorithms using mathematical constants
 - Methods:
 - `golden_ratio_optimization()` : Golden section search
 - `exponential_decay_schedule()` : Generate decay schedules
 - `adaptive_step_size()` : Calculate adaptive learning rates
4. **BoundedSystemAnalyzer**
 - Analyze constraints and system boundaries
 - Methods:
 - `analyze_constraint_boundaries()` : Identify optimal boundaries
 - `detect_emergent_patterns()` : Find patterns in time series

Dependencies: `numpy`, `scipy`, `math`

`constrained_intelligence/discovery.py`

Purpose: Discover mathematical constants from empirical data

Classes:

1. **DiscoveryMethods** (Enum)
 - Enumeration of discovery methods
 - Values: OPTIMIZATION_BASED, CONVERGENCE_ANALYSIS, PERIODICITY_DETECTION, etc.
2. **DiscoveryResult** (dataclass)
 - Stores discovery results
 - Includes confidence metrics and validation data
3. **ConstantDiscovery**
 - Main discovery engine
 - Methods:
 - `discover_from_optimization()` : Find constants in optimization trajectories
 - `detect_convergence_constants()` : Discover from convergence patterns
 - `discover_from_ratios()` : Analyze ratio sequences
 - `discover_from_boundaries()` : Find constants from boundary behavior
 - `validate_discovery()` : Validate discovered constants

Dependencies: numpy , scipy.stats , math**constrained_intelligence/constants.py****Purpose:** Define fundamental mathematical constants**Constants Defined:****Fundamental:**

- GOLDEN_RATIO : $\varphi \approx 1.618$
- EULER_NUMBER : $e \approx 2.718$
- PI : $\pi \approx 3.14159$

Derived:

- OPTIMAL_RESOURCE_SPLIT : $1/\varphi \approx 0.618$
- CONVERGENCE_THRESHOLD_FACTOR : $1/e \approx 0.368$
- LEARNING_RATE_BOUNDARY : $1/(2\pi) \approx 0.159$

System Boundaries:

- MAX EFFICIENCY_RATIO : 0.886
- MINIMAL_COMPLEXITY_CONSTANT : $e^{(1/e)} \approx 1.444$
- INFORMATION_DENSITY_LIMIT : $2 \cdot \ln(2) \approx 1.386$

Thresholds:

- HIGH_CONFIDENCE_THRESHOLD : 0.9
- MEDIUM_CONFIDENCE_THRESHOLD : 0.7
- VALIDATION_SIGNIFICANCE_LEVEL : 0.05

Helper Functions:

- `get_constant_info()` : Get details about a constant
- `list_all_constants()` : List all defined constants

Test Structure

`tests/test_core.py`

Tests for `core.py` module:

- Test `ConstantsMeasurement` methods
- Test `OptimizationEngine` algorithms
- Test `BoundedSystemAnalyzer` analysis functions
- Validate against known mathematical results

`tests/test_discovery.py`

Tests for `discovery.py` module:

- Test all discovery methods
- Validate discovered constants against theoretical values
- Test statistical validation
- Edge cases and error handling

`tests/test_constants.py`

Tests for `constants.py` module:

- Verify constant values
- Test helper functions
- Ensure mathematical relationships hold

Examples

`examples/basic_usage.py`

Simple, self-contained examples:

- Resource allocation
- Golden ratio optimization
- Learning rate schedules
- Convergence detection

`examples/advanced_examples.py`

Complex use cases:

- Multi-objective optimization
- Adaptive learning systems
- Real-world applications
- Performance comparisons

`examples/notebook.ipynb`

Interactive Jupyter notebook:

- Visualizations
- Step-by-step tutorials
- Experimental playground

Validation

`validation/experimental_validation.py`

Comprehensive validation suite:

- Validate against known mathematical results
- Empirical performance benchmarks
- Statistical significance tests
- Comparison with baseline methods

Functions:

- `validate_golden_ratio()` : Validate ϕ -based methods
- `validate_exponential_convergence()` : Test e-based predictions
- `run_all_validations()` : Execute full validation suite

Configuration Files

`setup.py`

Package configuration for PyPI:

- Package metadata (name, version, author)
- Dependencies
- Entry points
- Classifiers

`requirements.txt`

Runtime dependencies:

```
numpy>=1.19.0
scipy>=1.5.0
```

`requirements-dev.txt`

Development dependencies:

```
pytest>=6.0.0
pytest-cov>=2.10.0
black>=21.0
flake8>=3.8.0
mypy>=0.800
jupyter>=1.0.0
matplotlib>=3.3.0
```

`MANIFEST.in`

Additional files to include in distribution:

```
include README.md
include LICENSE
include requirements.txt
recursive-include examples *.py *.ipynb
recursive-include tests *.py
```

Docker

Dockerfile

Container definition:

- Base image: Python 3.8+
- Install package and dependencies
- Set up working directory
- Entry point for running examples/tests

.dockerignore

Exclude from Docker build:

- `__pycache__`
- `*.pyc`
- `.git`
- Virtual environments

CI/CD

.github/workflows/ci.yml

Automated testing and deployment:

- Run tests on push/PR
- Check code style
- Generate coverage reports
- Build and publish to PyPI (on release)

Jobs:

1. **test**: Run pytest on multiple Python versions
2. **lint**: Check code style with black and flake8
3. **type-check**: Run mypy type checker
4. **deploy**: Publish to PyPI (on tags)

Documentation

Markdown Files

- **README.md**: Main project documentation
- **QUICKSTART.md**: 5-minute getting started guide
- **THEORY.md**: Mathematical foundations
- **CONTRIBUTING.md**: Contribution guidelines
- **PACKAGE_STRUCTURE.md**: This file

Sphinx Documentation (Optional)

For generating HTML/PDF documentation:

```
cd docs/
make html
```

Development Workflow

1. Clone Repository

```
git clone https://github.com/yourusername/constrained-intelligence-constants.git
cd constrained-intelligence-constants
```

2. Install Development Environment

```
pip install -e ".[dev]"
```

3. Run Tests

```
pytest tests/
```

4. Run Examples

```
python examples/basic_usage.py
```

5. Run Validation

```
python validation/experimental_validation.py
```

Building and Distribution

Build Package

```
python setup.py sdist bdist_wheel
```

Install Locally

```
pip install -e .
```

Upload to PyPI

```
twine upload dist/*
```

Code Organization Principles

1. **Separation of Concerns:** Each module has a clear, single responsibility
2. **Modularity:** Components can be used independently
3. **Testability:** All functions/classes have corresponding tests
4. **Documentation:** Every public API is documented
5. **Examples:** Usage examples for all major features

Import Hierarchy

```

constants.py      # No internal imports (only stdlib/numpy)
↓
core.py          # Imports from constants.py
↓
discovery.py    # Imports from core.py and constants.py
↓
__init__.py      # Imports from all modules, defines public API

```

This prevents circular dependencies and maintains clean architecture.

File Naming Conventions

- **Modules:** lowercase_with_underscores.py
- **Classes:** CamelCase
- **Functions:** lowercase_with_underscores()
- **Constants:** UPPERCASE_WITH_UNDERSCORES
- **Tests:** test_*.py
- **Examples:** *_example.py or *_usage.py

Questions?

If you have questions about the package structure:

- Check this document
- Read the inline code comments
- Open a [GitHub Discussion](https://github.com/yourusername/constrained-intelligence-constants/discussions) (<https://github.com/yourusername/constrained-intelligence-constants/discussions>)

Maintained by the Constrained Intelligence Research Team