

Fox ML Infrastructure — System Architecture Diagram

This document provides a visual and textual representation of the Fox ML Infrastructure system architecture.

This architecture diagram is essential for enterprise technical reviews and integration planning.

1. Executive Summary

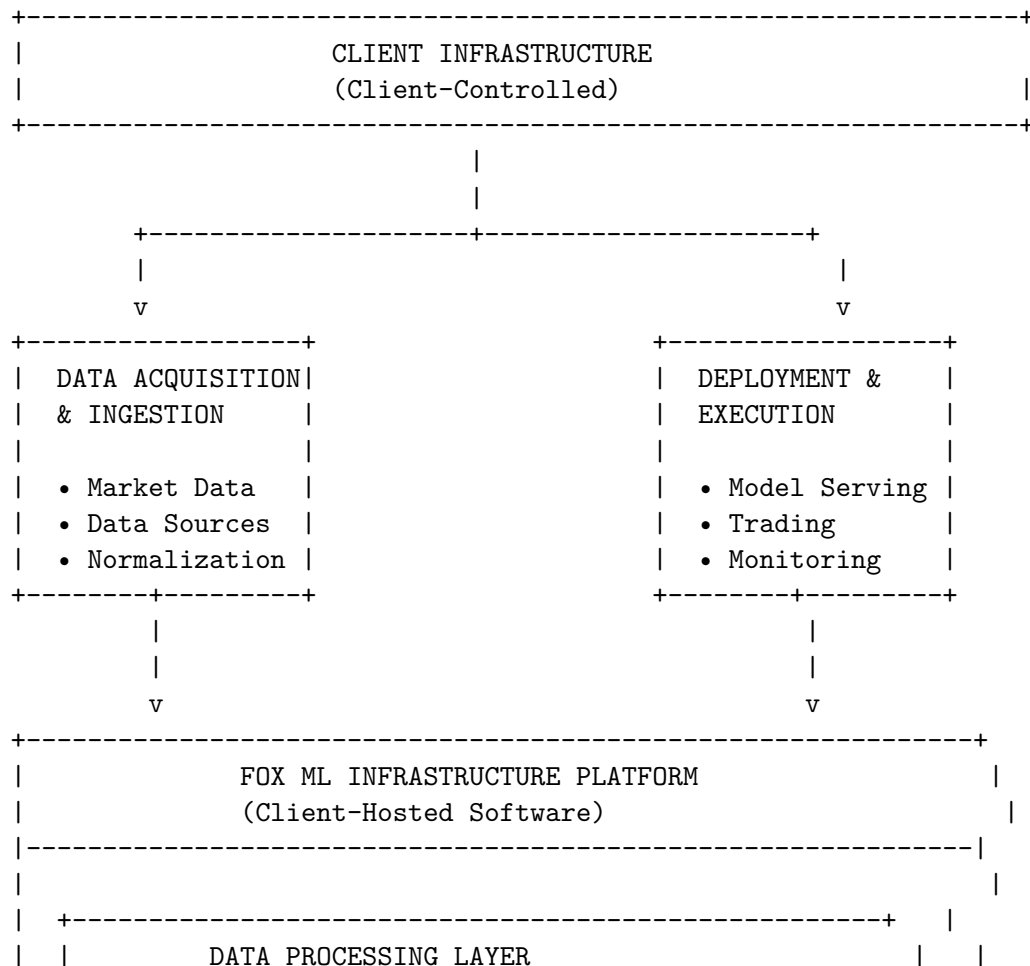
Fox ML Infrastructure is a client-hosted, modular ML research infrastructure platform.

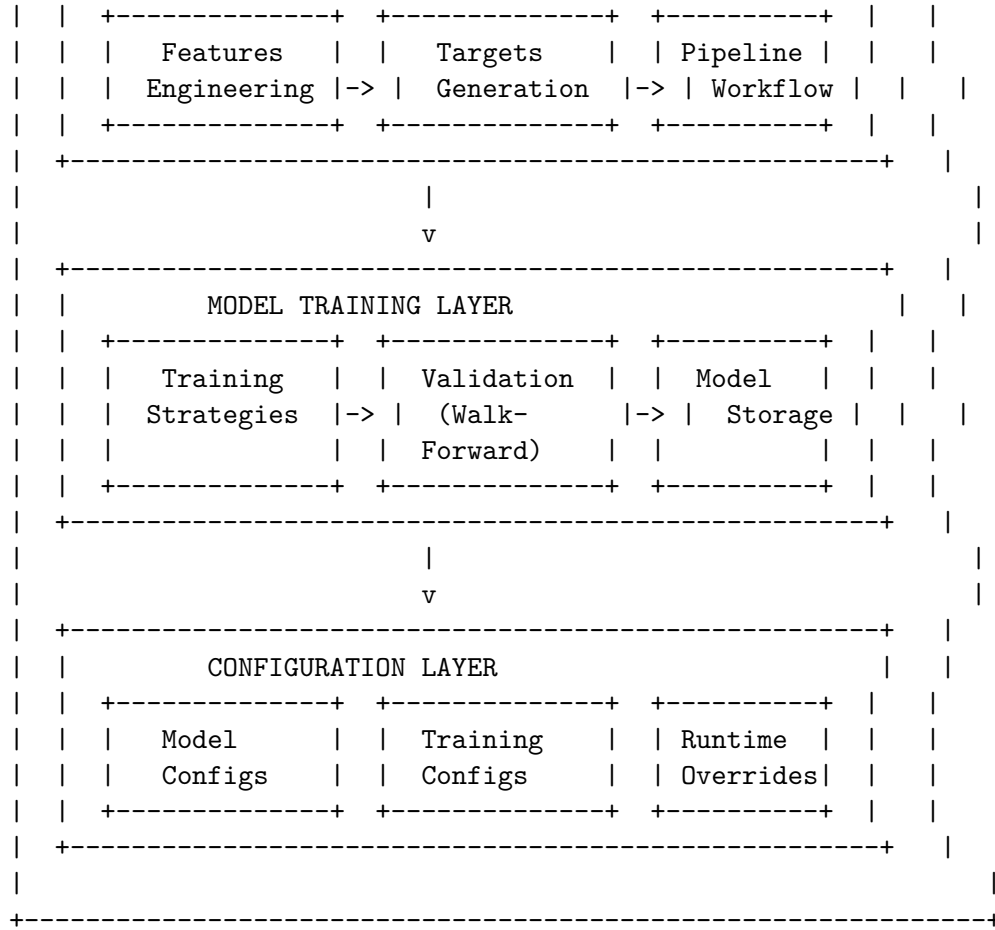
Key architectural principles:

- **Client-hosted** — Software runs entirely on client infrastructure
- **Modular design** — Clear separation of concerns across modules
- **Configuration-driven** — All runtime parameters from centralized configs
- **Pipeline-based** — Data flows through well-defined pipeline stages
- **Extensible** — Easy to add new models, features, and strategies

2. High-Level Architecture

2.1 System Overview





3. Component Architecture

3.1 Data Processing Layer

Purpose: Transform raw market data into ML-ready features and targets.

Components:

```

DATA_PROCESSING/
|-- features/
| |-- SimpleFeatureComputer      # Basic technical indicators
| |-- ComprehensiveFeatureBuilder # 200+ feature engineering
| +-- StreamingFeatureBuilder    # Streaming/online features
|-- targets/
| |-- BarrierTargetBuilder       # Barrier-based targets
| |-- ExcessReturnsBuilder       # Excess return targets
| +-- HFTForwardReturnsBuilder   # High-frequency targets
|-- pipeline/
| |-- DataNormalization         # RTH alignment, grid correction
| |-- FeaturePipeline           # End-to-end feature workflow
| +-- TargetPipeline            # End-to-end target workflow

```

```

+-- utils/
  |-- MemoryManagement      # Efficient memory handling
  |-- Logging                # Structured logging
  +-- Validation             # Data sanity checks

```

Data Flow:

```

Raw Market Data
  v
Normalization (RTH alignment, grid correction)
  v
Feature Engineering (200+ technical features)
  v
Target Generation (barrier, excess returns, HFT)
  v
Labeled Dataset (features + targets)

```

3.2 Model Training Layer

Purpose: Train and validate ML models using walk-forward validation.

Components:

```

TRAINING/
|-- model_fun/
|  |-- LightGBMTrainer      # Gradient boosting
|  |-- XGBoostTrainer      # Gradient boosting
|  |-- MLPTrainer          # Deep learning
|  |-- TransformerTrainer  # Transformer models
|  |-- EnsembleTrainer     # Model ensembles
|  |-- MultiTaskTrainer    # Multi-task learning
|  +-- [13+ additional trainers] # Probabilistic, advanced models
|-- strategies/
|  |-- SingleTaskStrategy  # One model per target
|  |-- MultiTaskStrategy  # Shared model for targets
|  +-- CascadeStrategy    # Sequential dependencies
|-- walkforward/
|  |-- WalkForwardValidator # Time-series validation
|  +-- PurgedTimeSeriesSplit # Leakage-safe splitting
+-- memory/
   |-- MemoryManager       # Memory optimization
   +-- BatchProcessing      # Efficient batch handling

```

Training Flow:

```

Labeled Dataset
  v
Walk-Forward Validation (time-series split)
  v

```

Model Training (17+ model types)
v
Model Validation (performance metrics)
v
Trained Models (serialized)

3.3 Configuration Layer

Purpose: Centralized, version-controlled configuration management.

Components:

```
CONFIG/  
|-- model_config/  
|   |-- lightgbm.yaml          # LightGBM configs (3 variants)  
|   |-- xgboost.yaml          # XGBoost configs (3 variants)  
|   |-- mlp.yaml              # MLP configs (3 variants)  
|   +-- [14+ additional configs] # All model types  
|-- training_config/  
|   |-- walkforward.yaml      # Walk-forward parameters  
|   |-- feature_selection.yaml # Feature selection configs  
|   +-- first_batch_specs.yaml # Training specifications  
+-- config_loader.py  
    |-- load_model_config()    # Load model configs  
    |-- list_available_configs() # List available configs  
    +-- apply_overrides()      # Runtime overrides
```

Configuration Flow:

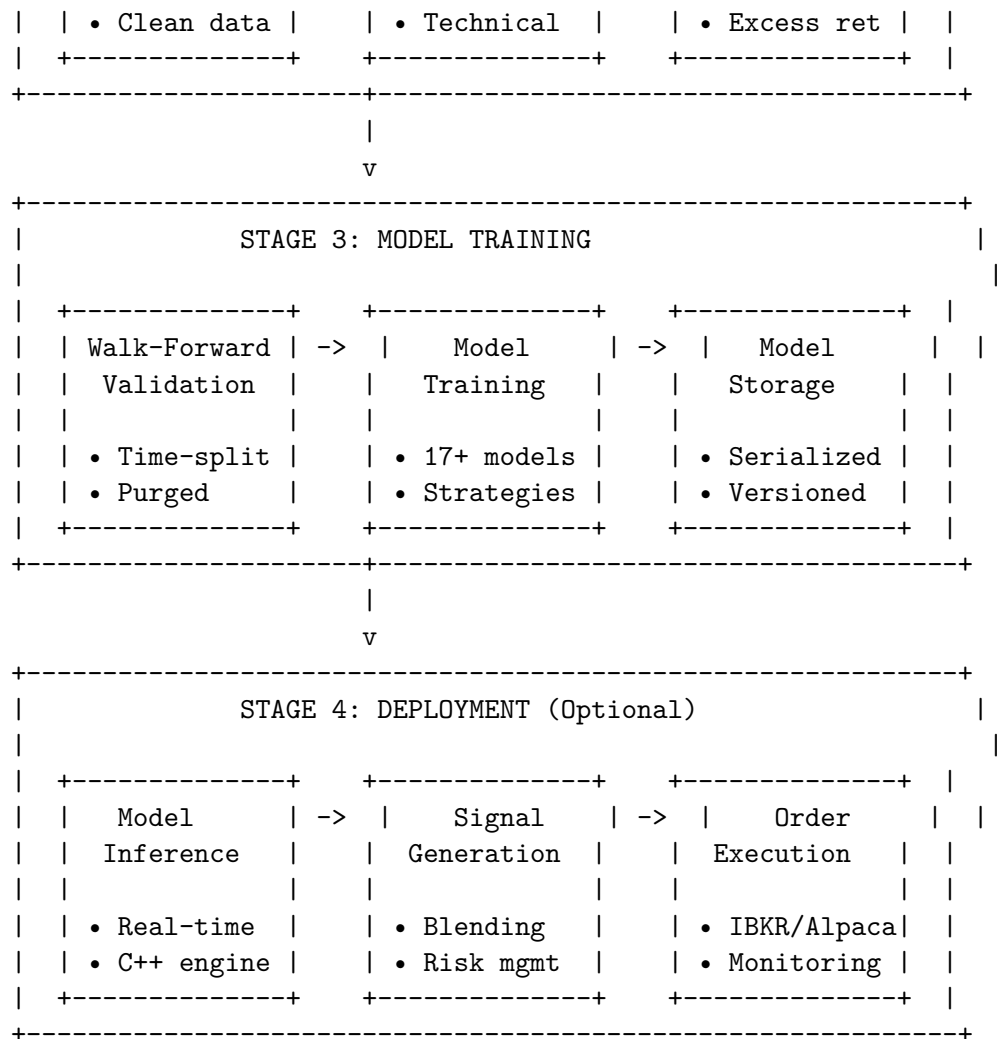
Base Config (YAML)
v
Variant Selection (conservative/balanced/aggressive)
v
Runtime Overrides (environment variables, CLI args)
v
Final Configuration (validated, applied)

3.4 Trading Integration Layer (Optional)

Purpose: Integration with trading brokers for paper and live trading.

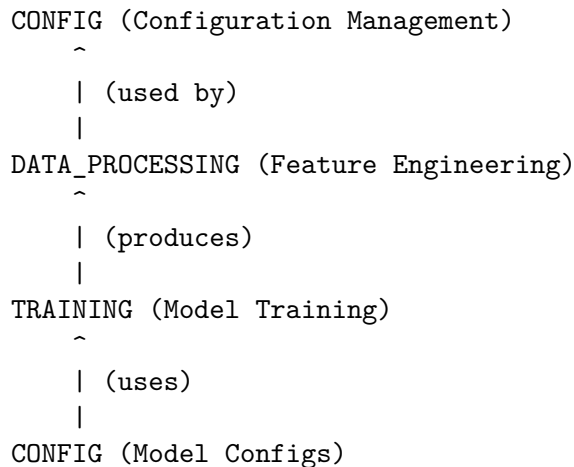
Components:

```
IBKR_trading/          # Interactive Brokers (Production)  
|-- live_trading/  
|   |-- MultiHorizonBlender # 5m, 10m, 15m, 30m, 60m blending  
|   |-- SafetyGuards        # Margin, short-sale, rate limiting  
|   |-- CostAwareArbitration # Cost-aware decision making
```

5. Module Dependencies

5.1 Dependency Graph



```

~
| (optional)
|
IBKR_trading / ALPACA_trading (Trading Integration)

```

5.2 External Dependencies

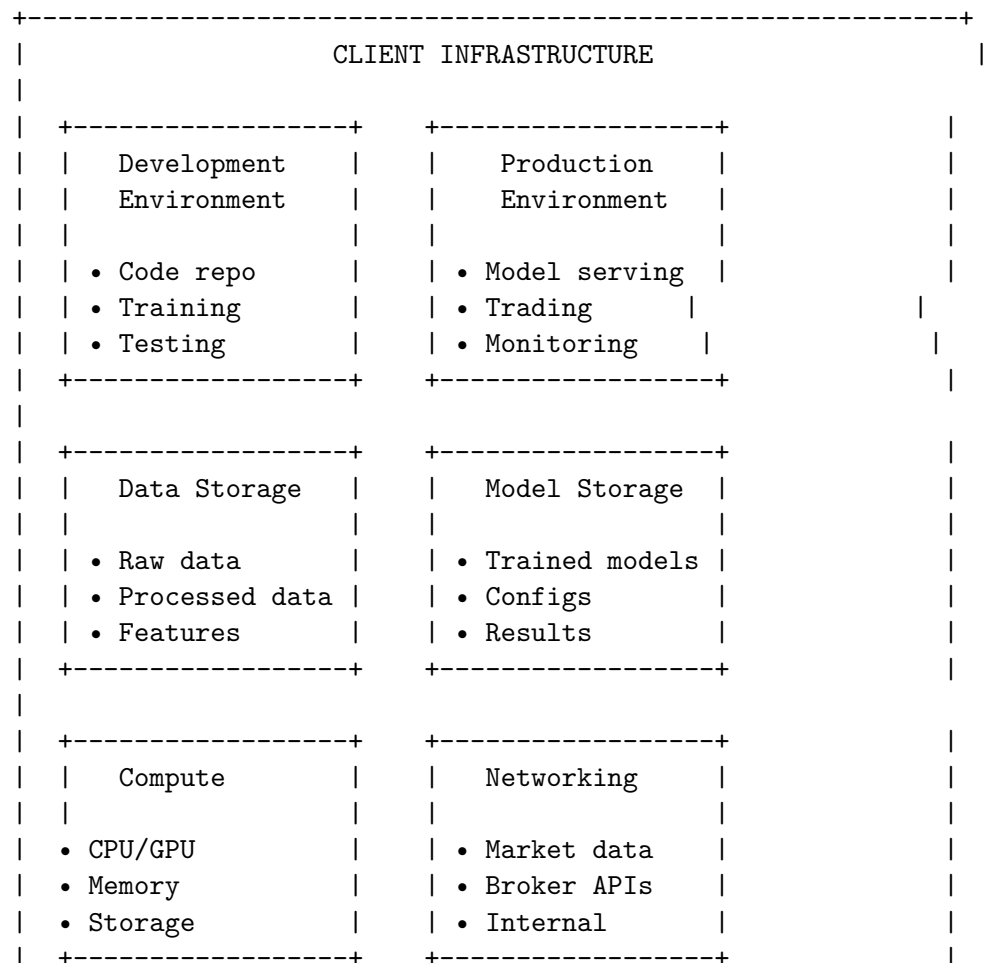
Core Dependencies: - **Python 3.9+** — Core runtime - **NumPy, Pandas** — Data processing - **Scikit-learn** — ML utilities - **LightGBM, XGBoost** — Gradient boosting models - **PyTorch** — Deep learning models (optional)

Trading Dependencies: - **ib_insync** — Interactive Brokers API - **alpaca-trade-api** — Alpaca API

Infrastructure: - **Git** — Version control - **GitHub** — Repository hosting - **YAML** — Configuration format

6. Deployment Architecture

6.1 Client-Hosted Deployment



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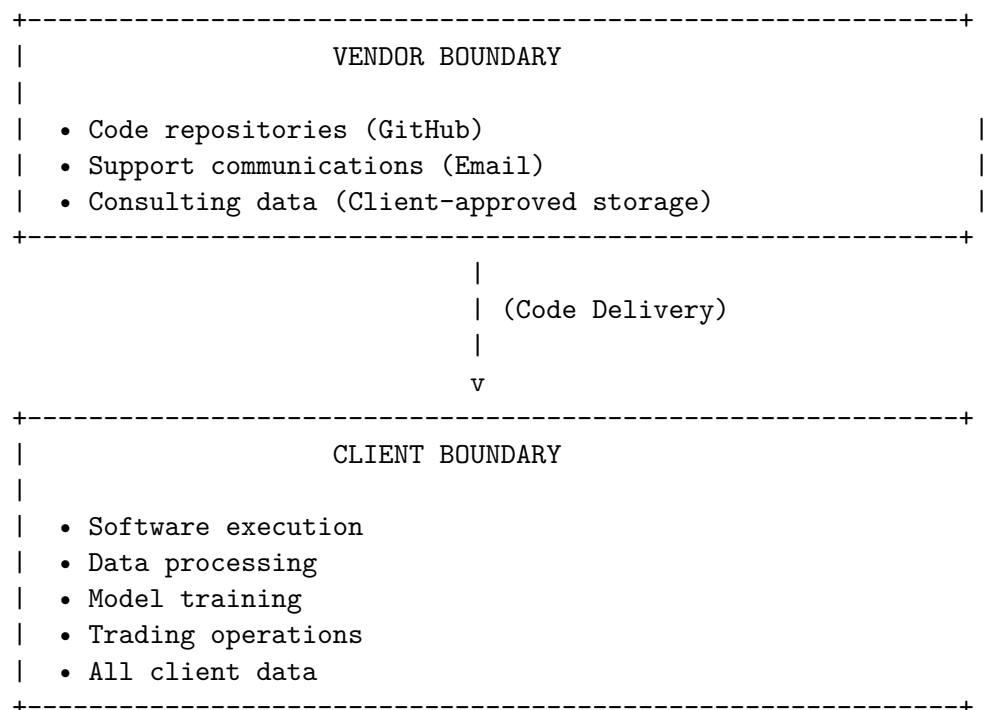
6.2 No Vendor Infrastructure

Important: Fox ML Infrastructure does not operate: - [X] Vendor-hosted servers - [X] Vendor-hosted databases - [X] Vendor cloud services - [X] Vendor network infrastructure

All infrastructure is client-controlled and client-managed.

7. Security Architecture

7.1 Security Boundaries



7.2 Security Controls

Vendor-side: - Repository access controls (2FA, SSH keys) - Code integrity (version control, signed commits) - Supply chain integrity (no telemetry, explicit dependencies)

Client-side: - Infrastructure security (client-managed) - Data security (client-managed) - Network security (client-managed) - Access control (client-managed)

8. Scalability Architecture

8.1 Horizontal Scalability

Scalable components: - **Data processing** — Parallel processing of multiple symbols/timeframes - **Model training** — Distributed training (if configured) - **Feature engineering** — Batch processing

for large datasets - **Inference** — Parallel inference for multiple models

8.2 Vertical Scalability

Resource requirements: - **CPU** — Multi-core processors (8+ cores recommended) - **RAM** — 16GB+ (32GB+ for large datasets) - **GPU** — Optional but recommended for deep learning - **Storage** — Sufficient for datasets and models

9. Integration Points

9.1 Data Integration

Data sources: - **yfinance** — Equity market data (default) - **Custom adapters** — Client-specific data sources - **File-based** — CSV, Parquet, HDF5 files

9.2 Model Integration

Model deployment: - **Serialized models** — Pickle, joblib, PyTorch formats - **C++ inference** — High-performance C++ inference engine (IBKR) - **REST API** — Model serving via REST API (if configured)

9.3 Trading Integration

Broker integrations: - **Interactive Brokers** — Production trading (multi-horizon, C++ engine) - **Alpaca** — Paper trading (regime-aware ensemble)

10. Configuration Architecture

10.1 Configuration Hierarchy

Base Config (YAML)
v
Variant Selection (conservative/balanced/aggressive)
v
Environment Overrides (environment variables)
v
Runtime Overrides (CLI arguments)
v
Final Configuration (validated, applied)

10.2 Configuration Sources

Configuration sources: - **YAML files** — CONFIG/model_config/*.yaml - **Environment variables** — Runtime overrides - **CLI arguments** — Command-line overrides - **Code defaults** — Fallback defaults

11. Monitoring and Observability

11.1 Logging

Logging components: - **Structured logging** — JSON-structured logs - **Log levels** — DEBUG, INFO, WARNING, ERROR - **Contextual information** — Run IDs, symbols, fold numbers - **Client-managed** — Logs stored and managed by client

11.2 Metrics

Metrics tracked: - **Performance metrics** — Training time, inference latency - **Model metrics** — Accuracy, loss, validation scores - **Pipeline metrics** — Data processing throughput - **Client-managed** — Metrics collected and stored by client

12. Contact

For architecture questions or integration planning:

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Subject: *Architecture Inquiry — Fox ML Infrastructure*

13. Related Documents

- docs/00_executive/ARCHITECTURE_OVERVIEW.md — Detailed architecture overview
 - docs/03_technical/design/ARCHITECTURE_DEEP_DIVE.md — Deep technical dive
 - LEGAL/ENTERPRISE_DELIVERY.md — Repository structure and delivery model
 - LEGAL/CLIENT_ONBOARDING.md — Client onboarding and integration guide
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14. Summary

Key Architectural Principles:

1. **Client-hosted** — Software runs entirely on client infrastructure
2. **Modular** — Clear separation of concerns across modules
3. **Configuration-driven** — All runtime parameters from configs
4. **Pipeline-based** — Data flows through well-defined stages
5. **Extensible** — Easy to add new models, features, strategies
6. **Scalable** — Horizontal and vertical scalability
7. **Secure** — Security boundaries and controls clearly defined

This architecture provides a comprehensive foundation for ML research and quantitative workflows.