Tading Bot Arquitectura y Especificaciones Finales - Trading Bot Ecosystem v2.0

Documento de Especificaciones Técnicas

Versión: 2.0

Fecha: Octubre 2025

Estado: 🗸 Actualizado - Multi-Timeframe + Rhai Support

Changelog: Multi-timeframe strategies, Rhai scripts, semantic constraints, warmup strategy

🎯 Visión General del Proyecto

Descripción

Sistema modular de trading algorítmico escrito en Rust que permite generar, testear y ejecutar miles de estrategias de trading **multi-timeframe** de forma automatizada, con arquitectura cliente-servidor basada en gRPC.

Objetivos Principales

- Generar 10,000+ estrategias automáticamente usando algoritmos genéticos **con constrainst semánticos**
- **Multi-timeframe strategies**: Combinar indicadores de diferentes timeframes (ej: EMA 1h + RSI 5m)
- Rhai scripts: Creación manual de estrategias con sintaxis amigable
- Backtest masivo de estrategias en minutos (modo Polars) vs realista (event-driven)
- ✓ Arquitectura cliente-servidor escalable
- Interface gráfica nativa moderna y rápida
- Warmup inteligente: Datos históricos automáticos para indicadores que requieren largos períodos

Casos de Uso

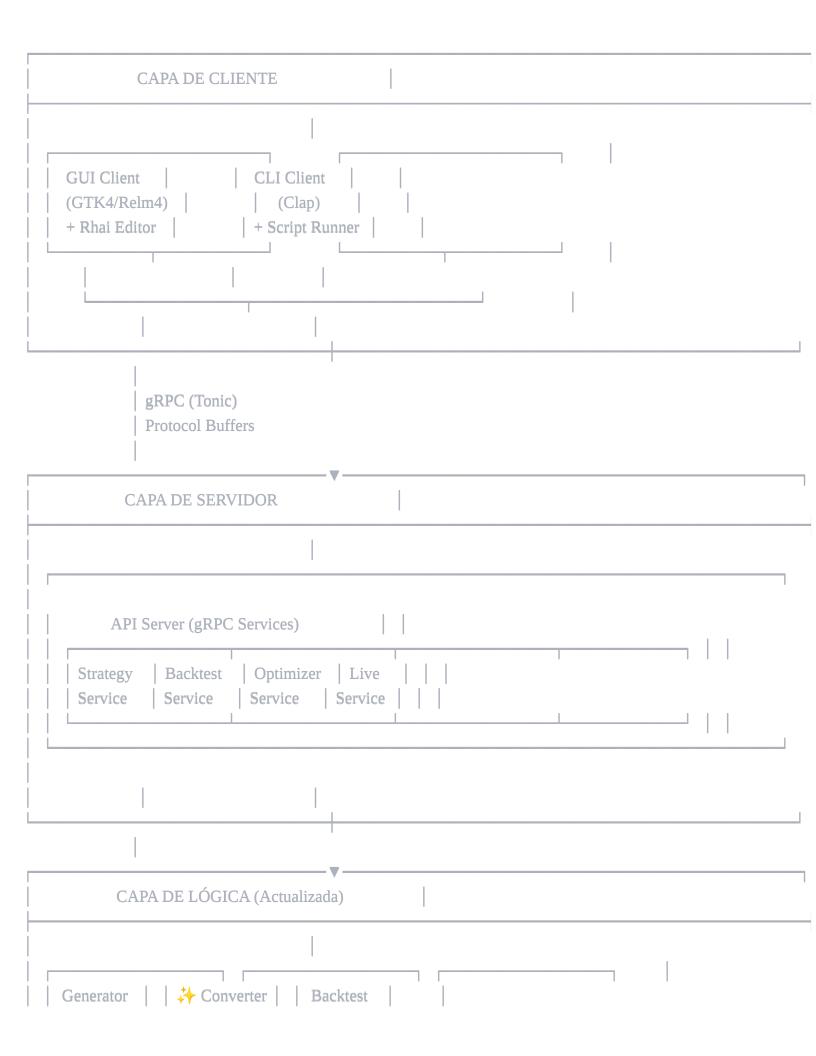
Usuario Workflow Herramientas Researcher/Quant Desarrollo y validación masiva Generador genético + Backtest Polars **Strategy Developer** Crear estrategias manuales Rhai scripts + StrategyBuilder Trader Algorítmico Ejecución automatizada 24/7 Event-driven backtest + Live runner

Portfolio Manager Gestión de múltiples estrategias GUI client + Multi-timeframe analysis

Arquitectura del Sistema

Diagrama de Alto Nivel Actualizado





(Genetic +	
Optimizer Runner Live Data Ma (Grid/GA) (Multi-TF + (Multi-T Warmup) Warmup)	
CAPA DE DATOS	
 Estrategias generadas (AST format) Estrategias manuales (Rhai format) Resultados de backtest Similarity scores pre-computados 	
Historical Data (Multi Timeframe Day	rquet)
Historical Data (Multi-Timeframe Par	
 - Precios OHLCV por timeframe - Warmup data cache - Indicator correlation matrix 	



📦 Estructura de Módulos (Crates) - Actualizada

Core Dependencies Actualizadas



```
core (types & traits)

    indicators (100% dinámico + metadata)

    data (multi-timeframe context)

 strategy-store (multi-format storage)
 strategy-generator (genetic + semantic constraints)
  – 거 strategy-converter (HUB central)
   inputs/ (rhai_parser, json_dsl, freqtrade)
    — outputs/ (polars_query, event_driven, rhai_runtime)
   similarity/ (correlation calculator)

    backtest-engine (dual: polars + event-driven)

 optimizer (grid/genetic/bayesian)
 runner-live (multi-TF + warmup)

    data-manager (multi-TF downloaders)

    api-proto (gRPC definitions)

 api-server (gRPC services)
 api-client (gRPC client)
— cli-client (terminal interface)
gui-client (GTK4 + Rhai editor)
```



🔧 Stack Tecnológico - Actualizaciones

Nuevas Adiciones

Componente	Tecnología	Versión	Justificación
Scripting	Rhai	1.19+	Estrategias manuales, sandboxed, hot reload
Correlación	nalgebra	0.32+	Cálculo de correlación Pearson entre indicadores
Multi-TF Synd	c Custom	-	Sincronización de timeframes con forward-fill
Script Editor	rGTK4 SourceVie	w5.0+	Syntax highlighting para Rhai

Mantenidos

- **Lenguaje**: Rust 2024 edition
- **Comunicación**: gRPC (Tonic) + Protocol Buffers

- **Data Processing**: Polars 0.41+ para backtest vectorizado
- GUI: GTK4/Relm4 para cliente nativo
- **Database**: SQLite (dev) → PostgreSQL (prod)



S Especificación Multi-Timeframe

1. Concepto: Timeframe Relativo Simple

Filosofía: En lugar de timeframes absolutos complejos, usar categorías semánticas:



rust

```
pub enum TimeframeCategory {
  Current, // Timeframe principal de la estrategia
  Medium, // 3-5x el timeframe principal
  High, // 12-24x el timeframe principal
```

2. Mapping Automático por Timeframe Principal

Principal Current Medium High				n Realismo			
1 m	1 m	5m	1h	V	Scalping + Context		
5m	5m	15m	1h	V	Day trading común		
15 m	15m	1h	4h	V	Swing trading		
1h	1h	4h	1d	V	Position trading		
4h	4h	1d	1w	V	Long-term		
1 d	1d	1w	1M	V	Investment		

3. Lógica de Evaluación



```
// Principio fundamental: Higher timeframes = vela cerrada anterior
// Current timeframe = vela actual en evaluación
impl MultiTimeframeStrategy {
         pub fn evaluate(&self, current_time: i64) -> Signal {
                  for condition in &self.entry_conditions {
                            let value = match condition.indicator.timeframe_category {
                                     TimeframeCategory::Current => {
                                            // West with the second of the
                                              self.get_current_indicator_value(&condition.indicator, current_time)
                                     TimeframeCategory::Medium | TimeframeCategory::High => {
                                            // Usar última vela CERRADA de timeframe superior
                                              self.get_last_closed_indicator_value(&condition.indicator, current_time)
                           };
                            if !self.evaluate_condition(condition, value) {
                                     return Signal::Hold;
                  Signal::Buy // Todas las condiciones cumplidas
```

4. Ejemplo Práctico

Strategy Timeline (Primary = 5m):



Especificación de Rhai Scripts

1. Sintaxis Propuesta

```
rust
```

```
// Ejemplo: Golden Cross + RSI Filter
// Configuración de timeframes
strategy_timeframe("5m"); // Timeframe principal
// Definición de indicadores con categorías
let ema_short = indicator("ema", [50], "current"); // EMA 50 en 5m
let ema_long = indicator("ema", [200], "medium"); // EMA 200 en 15m
let rsi = indicator("rsi", [14], "current"); // RSI 14 en 5m
let volume_avg = indicator("sma", [20], "high"); // Volume SMA en 1h
// Condiciones de entrada
entry_rules("and", [
  crosses_above(ema_short, ema_long), // Golden cross
  rsi < 50.0,
                        // RSI no sobbrecomprado
  volume() > volume_avg * 1.5
                            // Volumen alto
]);
// Condiciones de salida
exit_rules("or", [
  crosses_below(ema_short, ema_long), // Death cross
  rsi > 70.0,
                        // RSI sobrecomprado
  price() < ema_long * 0.98</pre>
                              // Stop loss 2%
1);
// Metadata opcional
set_name("Golden Cross RSI Filter");
set_description("EMA cross with RSI filter and volume confirmation");
```

2. Funciones Built-in

Función	Descripción	Ejemplo
strategy_timeframe(tf)	Define timeframe principa	lstrategy_timeframe("5m")
indicator(name, params, category)Crea indicador	<pre>indicator("rsi", [14], "current")</pre>
<pre>entry_rules(op, conditions)</pre>	Define reglas entrada	<pre>entry_rules("and", [])</pre>
<pre>exit_rules(op, conditions)</pre>	Define reglas salida	exit_rules("or", [])
crosses_above(a, b)	A cruza por encima de B	<pre>crosses_above(ema_fast, ema_slow)</pre>
<pre>crosses_below(a, b)</pre>	A cruza por debajo de B	<pre>crosses_below(price(), sma)</pre>
price()	Precio actual (close)	<pre>price() > sma</pre>
volume()	Volumen actual	<pre>volume() > avg_volume</pre>
set_name(name)	Nombre de estrategia	<pre>set_name("My Strategy")</pre>

3. Validaciones Automáticas



```
pub struct RhaiValidator;
impl RhaiValidator {
  pub fn validate_script(script: &str) -> Result<ValidationReport, RhaiError> {
    let mut report = ValidationReport::new();
    // Validar sintaxis Rhai
    let engine = rhai::Engine::new();
    engine.compile(script)?;
    // Validar semántica trading
    report.add_checks(vec![
       Self::check_timeframe_consistency(&script),
       Self::check_indicator_existence(&script),
       Self::check_logical_coherence(&script),
       Self::check_complexity_limits(&script),
    ]);
    Ok(report)
  fn check_timeframe_consistency(script: &str) -> ValidationCheck {
    // Verificar que todos los timeframes sean válidos
    // current/medium/high deben mapear a timeframes reales
  fn check_indicator_existence(script: &str) -> ValidationCheck {
    // Verificar que todos los indicadores existan en registry
  fn check_logical_coherence(script: &str) -> ValidationCheck {
    // Evitar condiciones contradictorias: rsi > 70 AND rsi < 30
```

Generador Masivo con Semantic Constraints

1. Constraints Dinámicos



```
pub struct SemanticConstraints {
  // Límites por categoría (dinámico)
  pub max_per_category: HashMap<IndicatorCategory, usize>,
  // Límite de similaridad (correlación)
  pub max_similarity_score: f64, // 0.0-1.0, default: 0.7
  // Límite de complejidad total
  pub max_complexity_score: f64,
  // Timeframes permitidos
  pub allowed_timeframe_categories: Vec<TimeframeCategory>,
impl Default for SemanticConstraints {
  fn default() -> Self {
    Self {
       max_per_category: hashmap! {
         IndicatorCategory::Trend => 2,
         IndicatorCategory::Momentum => 2,
         IndicatorCategory::Volume => 1,
         IndicatorCategory::Volatility => 1,
         IndicatorCategory::CandlestickPatterns => 1,
         IndicatorCategory::SupportResistance => 1,
         IndicatorCategory::Oscillators => 2,
       },
       max_similarity_score: 0.7, // 70% max correlation
       max_complexity_score: 15.0,
       allowed_timeframe_categories: vec![
         TimeframeCategory::Current,
         TimeframeCategory::Medium,
         TimeframeCategory::High,
       ],
```

2. Similarity Calculator (Correlación Real)



```
pub struct IndicatorSimilarityCalculator {
  // Cache de correlaciones pre-computadas
  correlation_matrix: HashMap<(String, String), f64>,
  reference_dataset: Vec<Candle>,
impl IndicatorSimilarityCalculator {
  pub fn calculate_similarity(&self, ind1: &str, ind2: &str) -> f64 {
    // Buscar en cache primero
     if let Some(cached) = self.get_cached_similarity(ind1, ind2) {
       return cached:
     }
    // Calcular correlación real usando datos históricos
     let values1 = self.calculate_indicator_series(ind1, &self.reference_dataset);
     let values2 = self.calculate_indicator_series(ind2, &self.reference_dataset);
    // Correlación de Pearson
     self.pearson_correlation(&values1, &values2).abs()
  fn pearson_correlation(&self, x: &[f64], y: &[f64]) -> f64 {
    // Implementación estándar de correlación Pearson
    // Resultado: -1.0 a 1.0 (usamos abs() para 0.0 a 1.0)
  pub fn build_correlation_matrix(&mut self) -> Result<(), Error> {
     let all_indicators = registry::all_names();
     for (i, ind1) in all_indicators.iter().enumerate() {
       for ind2 in all_indicators.iter().skip(i + 1) {
          let correlation = self.calculate_raw_correlation(ind1, ind2);
          self.correlation_matrix.insert((ind1.clone(), ind2.clone()), correlation);
    // Persistir cache en disco para futuras ejecuciones
     self.save_cache_to_disk()?;
     Ok(())
```

}

3. Generador Inteligente



```
impl RandomGenerator {
  pub fn generate_with_constraints(&self, constraints: &SemanticConstraints) -> StrategyAST {
    let mut selected_indicators = Vec::new();
    let mut category_count = HashMap::new();
    let mut complexity_score = 0.0;
    // Generate indicators respecting all constraints
    while selected_indicators.len() < self.max_indicators {
       let candidate = self.random_indicator_with_timeframe();
       let metadata = registry::get(&candidate.name).unwrap();
       // Check category limit
       let current_count = category_count.get(&metadata.category).unwrap_or(&0);
       let max_allowed = constraints.max_per_category.get(&metadata.category).unwrap_or(&999);
       if current_count >= max_allowed {
         continue;
       // Check complexity limit
       if complexity_score + metadata.complexity > constraints.max_complexity_score {
         continue:
       // Check similarity (most important)
       if self.is_too_similar(&candidate, &selected_indicators, constraints.max_similarity_score) {
         continue:
       }
      // Add indicator
       selected_indicators.push(candidate);
       *category_count.entry(metadata.category).or_insert(0) += 1;
       complexity_score += metadata.complexity;
    self.build_strategy_from_indicators(selected_indicators)
  fn random_indicator_with_timeframe(&self) -> IndicatorWithTimeframe {
```

```
let base_indicator = self.random_indicator();
  let timeframe_category = self.random_timeframe_category();
  IndicatorWithTimeframe {
    indicator: base_indicator,
    timeframe_category,
fn is_too_similar(&self, candidate: &IndicatorWithTimeframe, existing: &[IndicatorWithTimeframe], threshold: f64]
  existing.iter().any(|existing_ind| {
    let similarity = self.similarity_calculator.calculate_similarity(
       &candidate.indicator.name,
       &existing_ind.indicator.name,
    );
    similarity > threshold
  })
```



→ Backtest Engine Dual Mode

1. Arquitectura Dual



```
pub enum BacktestMode {
  VectorizedMassive, // Polars: 10,000+ strategies, fast approximation
  EventDrivenRealistic, // Event-by-event: 100 strategies, realistic simulation
pub struct DualBacktestEngine {
  polars_engine: PolarsBacktestEngine,
  event_engine: EventDrivenBacktestEngine,
  data_manager: MultiTimeframeDataManager,
impl DualBacktestEngine {
  pub async fn run_batch(&self, strategies: Vec<StrategyAST>, mode: BacktestMode) -> Vec<BacktestResult> {
    match mode {
       BacktestMode::VectorizedMassive => {
         // Convertir strategies a Polars queries
         let polars_queries: Vec<_> = strategies
            .iter()
            .map(|s| self.strategy_converter.to_polars_query(s))
            .collect();
         // Ejecutar en batch paralelo
         self.polars_engine.run_batch_parallel(polars_queries).await
       BacktestMode::EventDrivenRealistic => {
         // Convertir strategies a event-driven format
         let event_strategies: Vec<_> = strategies
            .iter()
            .map(|s| self.strategy_converter.to_event_driven(s))
            .collect();
         // Ejecutar secuencialmente con simulación completa
         let mut results = Vec::new();
         for strategy in event_strategies {
            let result = self.event_engine.run_single(strategy).await?;
            results.push(result);
         results
```

```
}
}
}
```

2. Polars Vectorized Engine



rust

```
pub struct PolarsBacktestEngine {
  // Optimizado para throughput masivo
impl PolarsBacktestEngine {
  pub async fn run_batch_parallel(&self, queries: Vec<PolarsBatchQuery>) -> Vec<BacktestResult> {
     use rayon::prelude::*;
    // Paralelizar por chunks
     queries.par_chunks(100)
       .flat_map(|chunk| {
         // Cada chunk se ejecuta en Polars
         self.execute_polars_chunk(chunk)
       })
       .collect()
  fn execute_polars_chunk(&self, queries: &[PolarsBatchQuery]) -> Vec<BacktestResult> {
    // Convertir estrategias a operations Polars
    // Ejecutar vectorizado con lazy evaluation
    // Retornar métricas básicas rápido
```

3. Event-Driven Engine



```
pub struct EventDrivenBacktestEngine {
  // Optimizado para realismo
  order_book: SimulatedOrderBook,
  slippage_model: SlippageModel,
  commission_model: CommissionModel,
impl EventDrivenBacktestEngine {
  pub async fn run_single(&self, strategy: EventDrivenStrategy) -> BacktestResult {
    let mut portfolio = Portfolio::new(10000.0);
    let mut trades = Vec::new();
    for candle in &self.data_manager.get_candles() {
       // Evaluar estrategia multi-timeframe
       let signal = strategy.evaluate_at_time(candle.timestamp, &self.data_manager);
       // Simular ejecución realista
       if let Some(order) = self.signal_to_order(signal, &portfolio) {
         let execution = self.order_book.simulate_execution(order, candle);
         let trade = self.apply_execution(execution, &mut portfolio);
         trades.push(trade);
    BacktestResult::from_trades(trades)
```

💾 Warmup Strategy para Live Trading

1. Límites Realistas por Timeframe



2. Warmup Decision Logic



```
pub enum WarmupAction {
  StreamingOnly,
                    // Para periods cortos (< 2 días)
  HistoricalDownload, // Para periods medios (2-365 días)
  StrategyNotViable, // Para periods largos (> 365 días)
}
impl LiveDataManager {
  pub async fn plan_warmup(&self, strategy: &StrategyAST) -> Result<WarmupPlan, Error> {
    let required_timeframes = strategy.required_timeframes();
    let mut plan = WarmupPlan::new();
    for tf in required_timeframes {
       let max_period = self.calculate_max_indicator_period(strategy, tf);
       let required_days = self.periods_to_days(max_period, tf);
       let limits = WarmupLimits::for_timeframe(tf);
       let action = if required_days <= limits.max_days_in_memory {</pre>
         WarmupAction::StreamingOnly
       } else if required_days <= limits.max_days_download {</pre>
         WarmupAction::HistoricalDownload
       } else {
         WarmupAction::StrategyNotViable
       };
       plan.add_timeframe(tf, action, required_days);
    }
    Ok(plan)
  fn calculate_max_indicator_period(&self, strategy: &StrategyAST, tf: TimeFrame) -> usize {
    let mut max_period = 0;
    // Buscar en todas las condiciones
    for condition in strategy.all_conditions() {
       if condition.indicator.timeframe_matches(tf) {
         let meta = registry::get(&condition.indicator.name).unwrap();
         // Calcular period basado en parámetros
```

```
// ej: EMA(200) = 200 periods, SMA(50) = 50 periods
let period = self.extract_period_from_params(&condition.indicator, meta);
    max_period = max_period.max(period);
}

// Factor de seguridad para warmup (1.5x)
(max_period as f64 * 1.5) as usize
}
```

3. Ejemplos de Warmup

Estrategia	Indicador Crítico	Time	frame Períodos	s Días	Acción
Scalping 1m	nRSI(14) 1m	1 m	14	0.01 🔽	Streaming
Day Trading	JEMA(50) 5m + RSI(14) 5m	m 5m	50	0.17 🔽	Streaming
Swing	EMA(200) 1h + $SMA(20)$	4h 1h	200	8.3 🔽	Download
Position	EMA(200) 1d	1d	200	200 🔽	Download
Long-term	EMA(500) 1d	1d	500	500 X	Not viable

Especificación gRPC Services - Actualizadas

1. StrategyService (Extendido)



protobuf

```
service StrategyService {
  // CRUD básico
  rpc List(ListRequest) returns (ListResponse);
  rpc Get(GetRequest) returns (Strategy);
  rpc Delete(DeleteRequest) returns (google.protobuf.Empty);
  // ** Nuevas operaciones
  rpc ParseRhai(ParseRhaiRequest) returns (ParseRhaiResponse);
  rpc ValidateStrategy(ValidateRequest) returns (ValidationResponse);
  rpc ConvertFormat(ConvertRequest) returns (ConvertResponse);
  // Operaciones pesadas (streaming)
  rpc Generate(GenerateRequest) returns (stream ProgressUpdate);
  rpc Export(ExportRequest) returns (stream ExportChunk);
// ** Nuevos mensajes
message ParseRhaiRequest {
  string rhai_script = 1;
  string target_timeframe = 2; // "5m", "1h", etc.
message ParseRhaiResponse {
  oneof result {
     StrategyAST strategy_ast = 1;
     ValidationError error = 2;
message ConvertRequest {
  StrategyAST strategy = 1;
  ConvertFormat target_format = 2;
enum ConvertFormat {
  POLARS_QUERY = 0;
  EVENT_DRIVEN = 1;
  RHAI_SCRIPT = 2;
```

```
FREQTRADE = 3;
}
```

2. BacktestService (Dual Mode)



protobuf

```
service BacktestService {
  // 🧩 Modo dual
  rpc RunMassive(MassiveBacktestRequest) returns (stream BacktestProgress);
  rpc RunRealistic(RealisticBacktestRequest) returns (stream BacktestProgress);
  // Multi-timeframe support
  rpc RunMultiTimeframe(MultiTimeframeBacktestRequest) returns (stream BacktestProgress);
  // Warmup planning
  rpc PlanWarmup(WarmupPlanRequest) returns (WarmupPlanResponse);
message MassiveBacktestRequest {
  repeated int64 strategy_ids = 1;
  string dataset = 2;
  TimeRange time_range = 3;
  PolarsConfig polars_config = 4; // Paralelización, chunks
message RealisticBacktestRequest {
  repeated int64 strategy_ids = 1; // Max 100 strategies
  string dataset = 2;
  TimeRange time_range = 3;
  SimulationConfig simulation = 4; // Slippage, commission, etc.
message WarmupPlanRequest {
  StrategyAST strategy = 1;
  repeated string required_timeframes = 2;
message WarmupPlanResponse {
  repeated WarmupAction actions = 1;
  bool is_viable = 2;
  string viability_reason = 3;
message WarmupAction {
  string timeframe = 1;
```

```
WarmupType type = 2;
 uint32 required_days = 3;
enum WarmupType {
 STREAMING_ONLY = 0;
 HISTORICAL_DOWNLOAD = 1;
 NOT_VIABLE = 2;
```



📊 Performance Benchmarks - Actualizados

1. Multi-Timeframe Performance

Operación	Input	Target	Notas
Rhai parsing	100 líneas script	< 50ms	Sintaxis validation
Strategy conversion	AST → Polars	< 100ms	Multi-TF compilation
Massive backtest	10,000 strategies	< 60 mir	Polars vectorized
Realistic backtest	100 strategies	< 30 mir	Event-driven simulation
Warmup planning	Multi-TF strategy	< 5ms	Decision logic
Historical download	1 year 1h data	< 2 min	Exchange API limits
Correlation calculation	100x100 indicators	s < 10 mir	One-time precompute

2. Memory Usage

```
Timeframe Max Period Data Size Memory
                                         Action
                    14 candles < 1KB <a>✓</a> Stream
         RSI(14)
                    200 candles < 10KB 🗸 Stream
5m
         EMA(200)
                    200 candles < 10KB ✓ Stream
1h
         EMA(200)
                    500 candles < 25KB ✓ Stream
1h
         EMA (500)
                    200 candles < 10KB 🗸 Stream
1d
         EMA(200)
1d
         EMA(500)
                    500 candles < 25KB ✓ Download
```



Testing Strategy - Multi-Timeframe

1. Unit Tests Extendidos



```
#[cfg(test)]
mod multi_timeframe_tests {
  use super::*;
  #[test]
  fn test_rhai_parsing_multi_timeframe() {
    let script = r#"
       strategy_timeframe("5m");
       let ema = indicator("ema", [200], "medium"); // Should map to 15m
       entry_rules("and", [ema > price()]);
    "#:
    let result = RhaiParser::parse(script).unwrap();
    assert_eq!(result.primary_timeframe, TimeFrame::M5);
    let ema_condition = &result.entry_rules.conditions[0];
    assert_eq!(ema_condition.indicator.timeframe_category, TimeframeCategory::Medium);
  #[test]
  fn test_semantic_constraints() {
    let constraints = SemanticConstraints::default();
    let generator = RandomGenerator::with_constraints(constraints);
    let strategy = generator.generate("Test".to_string());
    // Verificar que no hay indicadores muy correlacionados
    let indicators: Vec<_> = strategy.all_indicators().collect();
    for (i, ind1) in indicators.iter().enumerate() {
       for ind2 in indicators.iter().skip(i + 1) {
         let similarity = SIMILARITY_CALCULATOR.calculate_similarity(&ind1.name, &ind2.name);
         assert!(similarity < 0.7, "Indicators {} and {} too similar: {}", ind1.name, ind2.name, similarity);
  #[test]
  fn test_warmup_planning() {
    let strategy = create_test_multi_tf_strategy(); // EMA(200) 1h + RSI(14) 5m
```

```
let plan = LiveDataManager::plan_warmup(&strategy).unwrap();

assert_eq!(plan.actions.len(), 2);
assert_eq!(plan.actions[0].timeframe, TimeFrame::M5);
assert_eq!(plan.actions[0].action, WarmupAction::StreamingOnly);
assert_eq!(plan.actions[1].timeframe, TimeFrame::H1);
assert_eq!(plan.actions[1].action, WarmupAction::HistoricalDownload);
}
```

2. Integration Tests



```
#[tokio::test]
async fn test_full_multi_timeframe_workflow() {
  // 1. Create Rhai strategy
  let rhai_script = create_sample_multi_tf_script();
  // 2. Parse to AST
  let strategy_ast = RhaiParser::parse(&rhai_script).unwrap();
  // 3. Store in database
  let strategy_id = StrategyRepo::save(strategy_ast, "rhai").await.unwrap();
  // 4. Plan warmup
  let warmup_plan = LiveDataManager::plan_warmup(&strategy_ast).await.unwrap();
  assert!(warmup_plan.is_viable);
  // 5. Convert to backtest format
  let polars_query = StrategyConverter::to_polars_query(&strategy_ast).unwrap();
  // 6. Run backtest
  let result = DualBacktestEngine::run_single(polars_query, BacktestMode::VectorizedMassive).await.unwrap();
  // 7. Verify results
  assert!(result.total_trades > 0);
  assert!(result.sharpe_ratio.is_finite());
```

🔽 Roadmap Actualizado

Fase 1: Multi-Timeframe Foundation (Semanas 1-2)

- Completar data/multi_timeframe.rs con sincronización
- Extender StrategyAST para soportar timeframe categories
- **Implementar** TimeframeCategory mapping logic
- Testing multi-timeframe data loading

Fase 2: Rhai Integration (Semanas 3-4)

- Crear strategy-converter/inputs/rhai_parser.rs
- **V Implementar** Rhai → StrategyAST conversion
- Validación semántica de scripts Rhai
- 🔽 Testing Rhai parsing y validation

Fase 3: Semantic Constraints (Semanas 5-6)

- Calcular correlation matrix de indicadores
- V Implementar SemanticConstraints en generator
- **Testing** constraint enforcement
- **Optimizar** performance de similarity calculation

Fase 4: Backtest Engine Dual Mode (Semanas 7-10)

- Implementar Polars vectorized engine
- **V Implementar** Event-driven engine
- **V Strategy converter** outputs (AST → executable formats)
- **V Testing** ambos modos de backtest

Fase 5: Warmup Strategy (Semanas 11-12)

- **V Implementar** warmup planning logic
- **W Historical data** downloader para exchanges
- **Cache management** para datos históricos
- **Testing** warmup scenarios

Fase 6: API Layer (Semanas 13-14)

- **Extender** gRPC services con nuevas operaciones
- **Implementar** server-side handlers
- **Client integration** para multi-timeframe
- **Testing** end-to-end workflows

Fase 7: GUI Updates (Semanas 15-16)

- **Rhai editor** con syntax highlighting
- **W** Multi-timeframe visualization
- **Strategy builder** visual
- **Testing** user workflows

Success Metrics - Actualizados

Funcionalidad

Métrica	Target	Status		
Rhai scripts parseable	100% valid syntax	In	progress	
Multi-TF strategies generated	110,000+ combinations	In	progress	
Correlation accuracy	< 5% error vs real data	In	progress	
Warmup success rate	> 95% strategies viable	In	progress	
Backtest mode coverage	100% strategies both modes	In	progress	

Performance



Documentation Structure - Actualizada





⊚ Conclusión

Esta especificación v2.0 integra completamente:

✓ Multi-Timeframe Support

• Timeframe categories semánticas (current/medium/high)

- Evaluación realista (higher TF = vela cerrada)
- Mapping automático por timeframe principal

Rhai Scripting

- Sintaxis amigable para estrategias manuales
- · Validación semántica automática
- Conversión a StrategyAST unificado

Semantic Constraints

- Generador masivo inteligente
- Anti-correlación basada en datos reales
- Escalable a 100+ indicadores

Backtest Dual Mode

- Polars vectorizado: 10,000 strategies masivo
- Event-driven: 100 strategies realista
- Strategy converter como hub central

Warmup Strategy

- Límites realistas por timeframe
- Download automático vs streaming
- Validación de viabilidad de estrategias

La arquitectura mantiene la modularidad original pero agrega las capacidades avanzadas necesarias para un sistema de trading profesional multi-timeframe.

Próximo paso: Implementación fase por fase siguiendo el roadmap actualizado.

Documento preparado por: Trading Bot Team

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Estado: Aprobado para Desarrollo v2.0