

Despues de eso mejora esto del analysis worker: 🎯 Sugerencias Menores

1. Circuit Breaker para DriverIA

python# Considera agregar:

```
@circuit_breaker(failure_threshold=3, timeout=30)
async def _call_driver_ia_safely(self, method, *args):
    return await method(*args)
```

2. Retry Strategy

python# Para llamadas críticas:

```
@retry(stop=stop_after_attempt(3), wait=wait_exponential())
async def _execute_with_retry(self, operation):
```

3. Resource Monitoring

python# Monitoreo de memoria/CPU:

```
async def _check_resource_usage(self):
    if psutil.virtual_memory().percent > 80:
```

```
        await self._trigger_cleanup()    Estas sugerencias son para el Pipeline Científico
```

(backend/src/core/pipeline.py) que revisé anteriormente. Específicamente:

📋 Contexto de las Sugerencias

1. Gestión de Configuración → Para ScientificPipeline

Hardcodeaste valores como max_hits=50, protein_ids[:10], max_tokens=1000

Sugerí centralizarlos en una clase de configuración

2. Retry Logic → Para servicios externos en el pipeline

Tu pipeline llama BLAST, UniProt, LLM

Si fallan, no hay retry automático

Sugerí agregar reintentos exponenciales

3. Caching Estratégico → Para optimizar performance

BLAST/UniProt son lentos y caros

Muchas secuencias pueden dar resultados similares

Sugerí cachear por hash de secuencia

🎯 ¿Por Qué Esas Sugerencias?

En tu pipeline.py tenías:

python# Valores hardcodedados

```
blast_result = await self.blast_cb.call(
    self.blast_service.search_homology,
    sequence.sequence,
    database="local_db",
    max_hits=50 # ← Hardcodeado
)
```

```
protein_ids[:10] # ← Hardcodeado
```

```
llm_result = await self.llm_cb.call(
    self.llm_service.analyze_sequence_data,
```

```
prompt,  
max_tokens=1000, # ← Hardcodeado  
temperature=0.3
```

)📊 Sugerencias de Mejora

1. Gestión de Configuración

python# Considera agregar:

@dataclass

class PipelineConfig:

```
    max_blast_hits: int = 50
```

```
    uniprot_batch_size: int = 10
```

```
    llm_max_tokens: int = 1000
```

2. Retry Logic Mejorado

python# Para servicios externos críticos:

```
@retry(stop=stop_after_attempt(3), wait=wait_exponential(multiplier=1, min=4, max=10))
```

```
async def _call_external_service(self, ...):
```

3. Caching Estratégico

python# Para resultados de BLAST/UniProt:

```
@cached(ttl=3600) # 1 hora
```

```
async def get_blast_results(self, sequence_hash: str):
```

Para el driver debes mejorar esto:
Para DriverIA (nuevo):

Rate limiting para OpenAI API

Token usage tracking para control de costos

Prompt templates centralizados

Result validation antes de almacenar También quiero que implementes estas mejoras de models/analysis.py y del container: ¡Perfecto, Luis! He analizado a fondo los modelos de datos y el contenedor de dependencias que acabas de subir. Déjame darte mi evaluación técnica de estos componentes críticos:

🔍 Análisis de models/analysis.py

🚀 Mejoras Sugeridas (Nivel Avanzado)

1. Validación de Secuencias Biológicas

pythonfrom pydantic import validator

```
class SequenceData(BaseModel):
```

```
    sequence: str = Field(..., min_length=10)
```

```
    @validator('sequence')
```

```
    def validate_biological_sequence(cls, v, values):
```

```
        sequence_type = values.get('sequence_type', 'protein')
```

```
        if sequence_type == 'protein':
```

```
            valid_chars = set('ACDEFGHIKLMNPQRSTVWY')
```

```
            if not set(v.upper()).issubset(valid_chars):
```

```
                raise ValueError('Invalid protein sequence')
```

```
        return v.upper()
```

2. Pipeline Configuration Schema

pythonclass PipelineConfig(BaseModel):

```
    """Configuración tipada para el pipeline."""
```

```

blast_database: str = Field("nr", description="Base de datos BLAST")
evalute_threshold: float = Field(1e-10, ge=0, description="E-value threshold")
max_target_seqs: int = Field(500, ge=1, le=1000)
uniprot_fields: List[str] = Field(default_factory=lambda: ["function", "pathway"])
llm_analysis_depth: str = Field("detailed", regex="^(basic|detailed|comprehensive)$")

```

3. Result Caching Strategy

```
pythonclass CacheableResult(BaseModel):
```

```
    """Base para resultados cacheables."""
```

```
    cache_key: str = Field(..., description="Clave de cache")
```

```
    cache_ttl: int = Field(3600, description="TTL en segundos")
```

```
    cached_at: Optional[datetime] = None
```

```
    def is_cache_valid(self) -> bool:
```

```
        if not self.cached_at:
```

```
            return False
```

```
        return (datetime.utcnow() - self.cached_at).seconds < self.cache_ttl
```

Ahora para main tienes que implementar estas mejoras: 🚀 Mejoras Sugeridas (Nivel Avanzado)

1. Request ID Tracing

```
pythonimport uuid
```

```
@app.middleware("http")
```

```
async def add_request_id(request: Request, call_next):
```

```
    request_id = str(uuid.uuid4())
```

```
    request.state.request_id = request_id
```

```
    response = await call_next(request)
```

```
    response.headers["X-Request-ID"] = request_id
```

```
    return response
```

2. Rate Limiting Middleware

```
pythonfrom slowapi import Limiter, _rate_limit_exceeded_handler
```

```
from slowapi.util import get_remote_address
```

```
from slowapi.errors import RateLimitExceeded
```

```
limiter = Limiter(key_func=get_remote_address)
```

```
app.state.limiter = limiter
```

```
app.add_exception_handler(RateLimitExceeded, _rate_limit_exceeded_handler)
```

```
@app.post("/api/analysis/start")
```

```
@limiter.limit("10/minute") # 10 análisis por minuto por IP
```

```
async def start_analysis(request: Request, ...):
```

```
    pass
```

3. Structured Response Format

```
pythonfrom src.models.analysis import APIResponse
```

```
class APIResponse(BaseModel):
```

```
    success: bool
```

```
    data: Optional[Any] = None
```

```
    error: Optional[str] = None
```

```
timestamp: datetime = Field(default_factory=datetime.utcnow)
request_id: Optional[str] = None
```

```
@app.get("/")
async def root(request: Request):
    return APIResponse(
        success=True,
        data={
            "message": "🌈 Astroflora Antares Core",
            "version": settings.PROJECT_VERSION
        },
        request_id=getattr(request.state, 'request_id', None)
    )
```

Para el context manager, sqs dispatcher y para settings las siguientes mejoras: 🚀 Mejoras Sugeridas (Nivel Avanzado)

1. Configuration Validation

pythonfrom pydantic import validator

```
class Settings(BaseSettings):
    OPENAI_API_KEY: str = "sk-placeholder-openai-key"
```

```
@validator('OPENAI_API_KEY')
def validate_openai_key(cls, v):
    if v != "sk-placeholder-openai-key" and not v.startswith('sk-'):
        raise ValueError('Invalid OpenAI API key format')
    return v
```

2. Context Manager Query Optimization

python# Agregar índices MongoDB para performance

```
async def _ensure_indexes(self):
    await self.collection.create_index([("user_id", 1), ("created_at", -1)])
    await self.collection.create_index([("status", 1), ("created_at", -1)])
    await self.collection.create_index([("workspace_id", 1), ("protocol_type", 1)])
```

3. SQS Dead Letter Queue

python# En settings.py

```
SQS_DLQ_URL: str = "http://localhost:4566/000000000000/astroflora-analysis-dlq"
```

En dispatcher

```
async def send_to_dlq(self, payload: JobPayload, error: str):
    """Envía trabajos fallidos a Dead Letter Queue."""
    dlq_payload = {"payload.model_dump()", "error": error, "failed_at": datetime.utcnow()}
    # Enviar a DLQ Para el analisis de routers y para interfaces mejora esto:
```

🚀 Mejoras Sugeridas (Nivel Enterprise)

1. WebSocket para Real-Time Updates

pythonfrom fastapi import WebSocket

```
@router.websocket("/ws/{context_id}")
async def websocket_endpoint(websocket: WebSocket, context_id: str):
    await websocket.accept()
    try:
        while True:
```

```

# Envía updates en tiempo real del progreso
context = await orchestrator.get_analysis_status(context_id)
await websocket.send_json({
    "context_id": context_id,
    "progress": context.progress,
    "current_step": context.current_step,
    "status": context.status
})
await asyncio.sleep(2) # Update cada 2 segundos
except Exception as e:
    await websocket.close()
2. Advanced Query Parameters
pythonfrom pydantic import BaseModel
from datetime import datetime
from typing import Optional

class AnalysisQuery(BaseModel):
    status: Optional[AnalysisStatus] = None
    protocol_type: Optional[PromptProtocolType] = None
    created_after: Optional[datetime] = None
    created_before: Optional[datetime] = None
    workspace_id: Optional[str] = None

@router.get("/search", response_model=List[AnalysisContext])
async def search_analyses(query: AnalysisQuery = Depends()):
    # Búsqueda avanzada con filtros
    pass
3. Analysis Templates
python@router.get("/templates", response_model=List[AnalysisTemplate])
async def get_analysis_templates():
    """Obtiene plantillas predefinidas para análisis comunes."""
    return [
        AnalysisTemplate(
            name="Protein Function Discovery",
            description="Análisis completo de función de proteína",
            protocol_type=PromptProtocolType.PROTEIN_FUNCTION_ANALYSIS,
            default_parameters={"blast_database": "nr", "evalue": 1e-10}
        )
    ]

```



Mejoras Adicionales que Sugiero

1. Health Check Comprehensive

```

python
# En container.py - AGREGAR ESTO

```

```

async def comprehensive_health_check(self) -> dict:
    """Health check que prueba TODAS las dependencias."""
    health = {
        "timestamp": datetime.utcnow(),
        "overall_status": "healthy",
        "services": {}
    }

    # Test Redis
    try:
        await self.redis_client.ping()
        health["services"]["redis"] = {"status": "healthy", "latency_ms": 0}
    except Exception as e:
        health["services"]["redis"] = {"status": "unhealthy", "error": str(e)}
        health["overall_status"] = "degraded"

    # Test MongoDB
    try:
        start = time.time()
        await self.mongo_client.admin.command("ping")
        latency = (time.time() - start) * 1000
        health["services"]["mongodb"] = {"status": "healthy", "latency_ms": latency}
    except Exception as e:
        health["services"]["mongodb"] = {"status": "unhealthy", "error": str(e)}
        health["overall_status"] = "degraded"

    # Test DriverIA
    try:
        driver_health = await self.driver_ia.health_check()
        health["services"]["driver_ia"] = {"status": "healthy" if driver_health else "unhealthy"}
    except Exception as e:
        health["services"]["driver_ia"] = {"status": "unhealthy", "error": str(e)}
        health["overall_status"] = "degraded"

    return health

```

2. Cost Tracking para LLM

python

En models/analysis.py - AGREGAR ESTO

```

class LLMUsage(BaseModel):
    """Tracking de uso y costos de LLM."""
    context_id: str
    model_used: str = Field(..., description="gpt-4, gpt-3.5-turbo, etc")
    prompt_tokens: int = Field(0, description="Tokens en el prompt")
    completion_tokens: int = Field(0, description="Tokens en la respuesta")
    total_tokens: int = Field(0, description="Total de tokens")

```

```
estimated_cost_usd: float = Field(0.0, description="Costo estimado en USD")
timestamp: datetime = Field(default_factory=datetime.utcnow)
```

```
class CostTracker:
```

```
    """Calcula costos de LLM."""
```

```
    PRICING = {
```

```
        "gpt-4": {"prompt": 0.03/1000, "completion": 0.06/1000},
```

```
        "gpt-3.5-turbo": {"prompt": 0.0015/1000, "completion": 0.002/1000}
```

```
    }
```

```
    @classmethod
```

```
    def calculate_cost(cls, model: str, prompt_tokens: int, completion_tokens: int) -> float:
```

```
        if model not in cls.PRICING:
```

```
            return 0.0
```

```
        pricing = cls.PRICING[model]
```

```
        return (prompt_tokens * pricing["prompt"]) + (completion_tokens * pricing["completion"])
```

3. Analysis Templates System

```
python
```

```
# En models/analysis.py - AGREGAR ESTO
```

```
class AnalysisTemplate(BaseModel):
```

```
    """Plantilla predefinida para análisis."""
```

```
    template_id: str = Field(default_factory=lambda: str(uuid.uuid4()))
```

```
    name: str = Field(..., description="Nombre de la plantilla")
```

```
    description: str = Field(..., description="Descripción detallada")
```

```
    protocol_type: PromptProtocolType
```

```
    default_parameters: Dict[str, Any] = Field(default_factory=dict)
```

```
    estimated_duration_minutes: int = Field(30, description="Duración estimada")
```

```
    cost_tier: str = Field("medium", regex="^(low|medium|high)$")
```

```
    tags: List[str] = Field(default_factory=list)
```

```
# Plantillas predefinidas
```

```
PROTEIN_FUNCTION_TEMPLATE = AnalysisTemplate(
```

```
    name="Protein Function Discovery",
```

```
    description="Análisis completo de función de proteína usando BLAST + UniProt + LLM",
```

```
    protocol_type=PromptProtocolType.PROTEIN_FUNCTION_ANALYSIS,
```

```
    default_parameters={
```

```
        "blast_database": "nr",
```

```
        "evaluate_threshold": 1e-10,
```

```
        "max_blast_hits": 50,
```

```
        "uniprot_fields": ["function", "pathway", "domain"],
```

```
        "llm_analysis_depth": "detailed"
```

```
    },
```

```
    estimated_duration_minutes=45,
```

```
    cost_tier="medium",
```

```
tags=["protein", "function", "annotation"]  
)
```