# **Integration Guide**

This guide provides detailed instructions for integrating the VMware MCP Server with Ollama (local LLM) and n8n (workflow automation).

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# **Ollama Integration**

## **Prerequisites**

1. Install Ollama

```bash

# Linux/macOS

curl -fsSL https://ollama.ai/install.sh | sh

# Or download from https://ollama.ai/download

#### 1. Pull Required Models

```bash

# Pull recommended model ollama pull llama3.2

# Alternative models

ollama pull codellama

ollama pull mistral

ollama pull gemma2

` ` `

#### 1. Start Ollama Service

```bash

# Start Ollama daemon

ollama serve

# Verify it's running

curl http://localhost:11434/api/tags

` ` `

## Configuration

#### 1. Environment Variables

```bash

# Enable Ollama integration export ENABLE OLLAMA=true

```
# Configure Ollama endpoint
export OLLAMA HOST=http://localhost:11434
# Set default model
export OLLAMA MODEL=Ilama3.2
# Set timeout
export OLLAMA_TIMEOUT=30
 1. Docker Compose Setup
   ```yaml
   # Already included in docker-compose.yml
   image: ollama/ollama:latest
   container name: ollama
   ports:
     · "11434:11434"
      volumes:
     ollama data:/root/.ollama
       restart: unless-stopped
Usage Examples
 1. VM Performance Analysis
   ```python
   # Using MCP tool
   result = await mcp_server.call_tool("analyze_vm_performance", {
   "vm name": "web-server-01",
   "user_role": "operator"
   })
print(result)
# Output: Al analysis of VM performance with recommendations
 1. VM Sizing Recommendations
    python
   result = await mcp_server.call_tool("suggest_vm_sizing", {
          "workload_description": "High-traffic web application with database",
   "requirements": {
              "expected_users": 10000,
              "peak_traffic": "5000 concurrent users",
    "database_size": "500GB",
   "availability": "99.9%"
   }
   })
 2. Troubleshooting Assistance
    python
```

result = await mcp\_server.call\_tool("troubleshoot\_issue", {

"issue\_description": "VM is running slowly and consuming high CPU",

```
"vm_name": "app-server-02"
})
```

#### **Custom AI Workflows**

#### 1. Create Custom Analysis

```
```python
from src.ollama_integration import ollama_integration
async def custom_capacity_analysis(cluster_data):
prompt = f"""
```

Analyze this VMware cluster capacity data and predict:

- 1. When resources will be exhausted
- 2. Recommended scaling actions
- 3. Cost optimization opportunities

```
Data: {cluster_data}
"""

return await ollama_integration.generate_response(prompt, {
    "type": "capacity_planning",
    "data": cluster_data
})
```

. . .

#### 1. Streaming Responses

```
python
  async def stream_troubleshooting_steps(issue):
    async for chunk in ollama_integration.stream_response(
        f"Provide step-by-step troubleshooting for: {issue}"
    ):
        print(chunk, end="", flush=True)
```

# n8n Integration

### **Prerequisites**

```
1. Install n8n
```

```
""bash
# Using npm
npm install n8n -g

# Using Docker
docker run -it -rm -name n8n -p 5678:5678 n8nio/n8n

# Using Docker Compose (included in project)
docker-compose up n8n

""
```

#### 1. Access n8n Interface

- Open http://localhost:5678
- Create admin account
- Configure basic settings

# **Configuration**

#### 1. Environment Variables

```
"``bash
# Enable n8n integration
export ENABLE_N8N=true

# Configure webhook URL
export N8N_WEBHOOK_URL=http://localhost:5678/webhook

# Set API key (if using authentication)
export N8N_API_KEY=your-api-key-here
```

#### 1. Webhook Setup in n8n

```
json
{
"nodes": [
{
"parameters": {
"httpMethod": "POST",
"path": "vmware-events",
"options": {}
},
   "name": "Webhook",
   "type": "n8n-nodes-base.webhook",
"typeVersion": 1,
"position": [250, 300]
}
]
}
```

# **Workflow Examples**

### 1. VM Lifecycle Management

```
}
"name": "Check Event Type",
"type": "n8n-nodes-base.if"
},
{
"parameters": {
"to": "admin@company.com",
   "subject": "New VM Created: {{$json[\"vm_data\"][\"name\"]}}",
"text": "A new VM has been created with the following details:\n\nName:
{{$json[\"vm_data\"][\"name\"]}}\nCPU: {{$json[\"vm_data\"][\"cpu_count\"]}}\nMemory:
{{$json[\"vm_data\"][\"memory_mb\"]}}MB"
"name": "Send Email Notification",
"type": "n8n-nodes-base.emailSend"
}
]
}
```

#### 2. Performance Monitoring

```
json
{
"name": "Performance Alert Workflow",
"nodes": [
{
"parameters": {
"path": "performance-alert"
},
"name": "Performance Webhook",
"type": "n8n-nodes-base.webhook"
},
"parameters": {
  "conditions": {
 "number": [
     "value1": "={{$json[\"metrics\"][\"cpu_utilization_percent\"]}}",
    "operation": "larger",
   "value2": 80
"name": "Check CPU Threshold",
"type": "n8n-nodes-base.if"
},
"parameters": {
```

#### 3. Maintenance Automation

```
json
{
"name": "Maintenance Window Automation",
"nodes": [
{
"parameters": {
"path": "maintenance-window"
},
"name": "Maintenance Webhook",
"type": "n8n-nodes-base.webhook"
},
{
"parameters": {
"functionCode": "// Create maintenance tasks\nconst tasks = [\n { action: 'mi-
grate_vms', priority: 1 },\n { action: 'enter_maintenance', priority: 2 },\n { action:
'apply_patches', priority: 3 },\n { action: 'exit_maintenance', priority: 4 }\n];\n\nreturn
tasks.map(task => ({ json: task }));"
"name": "Generate Tasks",
"type": "n8n-nodes-base.function"
},
{
 "parameters": {
   "url": "http://vmware-mcp-server:8080/api/tools/{{$json[\"action\"]}}",
  "options": {
   "headers": {
   "Authorization": "Bearer {{$env[\"MCP_API_TOKEN\"]}}"
}
"name": "Execute VMware Action",
"type": "n8n-nodes-base.httpRequest"
}
]
}
```

## **Integration Usage**

```
1. Send Events from VMware MCP Server
```

```
```python
   from src.n8n_integration import n8n_integration
# Send VM event
await n8n_integration.send_vm_event("vm_created", {
"name": "new-web-server",
"cpu count": 4,
"memory mb": 8192,
"created by": "admin"
})
# Send performance alert
await n8n integration.send alert(
"high_cpu_usage",
"warning",
"CPU usage exceeded 80%",
{"vm name": "web-server-01", "cpu percent": 85}
. . .
```

### 1. Custom Webhook Endpoints

```
python
  # Send to custom webhook
  await n8n_integration.send_custom_webhook("backup-status", {
        "backup_type": "vm_snapshot",
        "status": "completed",
        "vm_list": ["web-01", "db-01", "app-01"],
        "duration": "45 minutes"
})
```

# **Combined Workflows**

#### **AI-Powered Automation**

### 1. Intelligent Performance Optimization

```
"" python
async def intelligent_performance_workflow(vm_name):
# Get VM performance data
vm_data = await vm_ops.get_vm_resource_usage(vm_name)

# Get Al analysis
ai_analysis = await ollama_integration.analyze_vm_performance(vm_data)

# Send to n8n for automated actions
await n8n_integration.trigger_workflow("performance_optimization", {
    "vm_name": vm_name,
    "performance_data": vm_data,
    "ai_recommendations": ai_analysis,
    "automation level": "supervised" # or "automatic"
```

})

#### 2. Predictive Maintenance

```
```python
async def predictive_maintenance_workflow():
# Get cluster resources
cluster_data = await resource_ops.get_cluster_resources()
# AI analysis for capacity planning
capacity_analysis = await ollama_integration.generate_response(
"Analyze cluster capacity and predict maintenance needs",
{"cluster_data": cluster_data}
# Trigger maintenance planning workflow
await n8n_integration.send_maintenance_notification(
"predictive maintenance",
["cluster-01"],
{
"predicted_date": "2024-02-15",
"ai_analysis": capacity_analysis,
"confidence": "high"
}
)
```

## **Example n8n Workflow with AI Integration**

```
{
  "name": "AI-Powered VM Optimization",
  "nodes": [
    {
      "parameters": {
        "path": "ai-optimization"
      "name": "Optimization Webhook",
      "type": "n8n-nodes-base.webhook"
    },
    {
      "parameters": {
        "url": "http://ollama:11434/api/generate",
        "options": {
          "headers": {
            "Content-Type": "application/json"
          "body": {
            "model": "llama3.2",
            "prompt": "Analyze VM performance: {{$json[\"vm_data\"]}} and suggest op-
timizations",
            "stream": false
          }
        }
      },
      "name": "Get AI Analysis",
      "type": "n8n-nodes-base.httpRequest"
    },
      "parameters": {
        "functionCode": "// Parse AI response and create action plan\nconst aiResponse
= JSON.parse($input.first().json.body).response;\nconst actions = [];\n\nif (aiRe-
sponse.includes('increase memory')) {\n actions.push({type: 'modify_resources',
memory_increase: true});\n\nif (aiResponse.includes('reduce CPU')) {\n ac-
tions.push({type: 'modify_resources', cpu_reduce: true});\n}\n\nreturn
actions.map(action => ({json: action}));"
      "name": "Parse AI Recommendations",
      "type": "n8n-nodes-base.function"
    },
      "parameters": {
        "url": "http://vmware-mcp-server:8080/api/tools/modify_vm_resources",
        "options": {
          "headers": {
            "Authorization": "Bearer {{$env[\"MCP_TOKEN\"]}}"
          "body": {
            "vm_name": "{{$json[\"vm_name\"]}}",
            "cpu_count": "{{$json[\"new_cpu_count\"]}}",
            "memory_mb": "{{$json[\"new_memory_mb\"]}}"
          }
        }
      },
      "name": "Apply Optimizations",
      "type": "n8n-nodes-base.httpRequest"
 ]
}
```

# **Troubleshooting**

#### **Common Issues**

```
1. Ollama Connection Issues
   ```bash
   # Check if Ollama is running
   curl http://localhost:11434/api/tags
# Check logs
docker logs ollama
# Restart service
docker-compose restart ollama
 1. n8n Webhook Issues
   ```bash
   # Test webhook endpoint
   curl -X POST http://localhost:5678/webhook/test \
   -H "Content-Type: application/json" \
   -d '{"test": "data"}'
# Check n8n logs
docker logs n8n
 1. Integration Health Checks
   ```python
   # Check integration status
   ollama status = await ollama integration.health check()
   n8n_status = await n8n_integration.health_check()
print(f"Ollama: {ollama status}")
print(f"n8n: {n8n_status}")
Performance Optimization
 1. Ollama Performance
   ```bash
   # Use GPU acceleration (if available)
   docker run -gpus all ollama/ollama
# Optimize model loading
export OLLAMA_NUM_PARALLEL=4
export OLLAMA MAX LOADED MODELS=2
 1. n8n Performance
   ```bash
   # Increase worker processes
   export N8N_WORKERS=4
# Configure database for better performance
export DB_TYPE=postgresdb
```

```
export DB_POSTGRESDB_HOST=postgres
```

# **Monitoring and Logging**

#### 1. Enable Debug Logging

```
bash
```

```
export LOG_LEVEL=DEBUG
export ENABLE_AUDIT_LOG=true
```

#### 2. Monitor Integration Metrics

```
```python
```

# Custom metrics collection

from prometheus\_client import Counter, Histogram

```
ollama_requests = Counter('ollama_requests_total', 'Total Ollama requests')
n8n_webhooks = Counter('n8n_webhooks_total', 'Total n8n webhooks')
```

# **Best Practices**

#### 1. Security

- Use API keys for n8n authentication
- Implement rate limiting for AI requests
- Validate all webhook payloads

#### 2. Performance

- Cache AI responses for similar queries
- Use async operations for all integrations
- Implement circuit breakers for external services

#### 3. Reliability

- Implement retry logic with exponential backoff
- Use health checks to monitor service availability
- Set appropriate timeouts for all operations

#### 4. Monitoring

- Track integration usage metrics
- Monitor response times and error rates
- Set up alerts for service failures

For additional support, refer to the main documentation or contact the development team.