```
import json
import logging
import time
import uuid
from datetime import datetime
from typing import Any, Dict, List, Optional
import yaml
from pydantic import BaseModel
from swarm_models.tiktoken_wrapper import TikTokenizer
logger = logging.getLogger(__name__)
class MemoryMetadata(BaseModel):
  """Metadata for memory entries"""
  timestamp: Optional[float] = time.time()
  role: Optional[str] = None
  agent_name: Optional[str] = None
  session_id: Optional[str] = None
  memory_type: Optional[str] = None # 'short_term' or 'long_term'
  token_count: Optional[int] = None
  message_id: Optional[str] = str(uuid.uuid4())
```

class MemoryEntry(BaseModel): """Single memory entry with content and metadata""" content: Optional[str] = None metadata: Optional[MemoryMetadata] = None class MemoryConfig(BaseModel): """Configuration for memory manager""" max_short_term_tokens: Optional[int] = 4096 max_entries: Optional[int] = None system_messages_token_buffer: Optional[int] = 1000 enable_long_term_memory: Optional[bool] = False auto_archive: Optional[bool] = True archive_threshold: Optional[float] = 0.8 # Archive when 80% full class MemoryManager: 11 11 11 Manages both short-term and long-term memory for an agent, handling token limits, archival, and context retrieval. Args: config (MemoryConfig): Configuration for memory management

tokenizer (Optional[Any]): Tokenizer to use for token counting

```
def __init__(
  self,
  config: MemoryConfig,
  tokenizer: Optional[Any] = None,
  long_term_memory: Optional[Any] = None,
):
  self.config = config
  self.tokenizer = tokenizer or TikTokenizer()
  self.long_term_memory = long_term_memory
  # Initialize memories
  self.short_term_memory: List[MemoryEntry] = []
  self.system_messages: List[MemoryEntry] = []
  # Memory statistics
  self.total_tokens_processed: int = 0
  self.archived_entries_count: int = 0
def create_memory_entry(
  self,
  content: str,
  role: str,
  agent_name: str,
```

long_term_memory (Optional[Any]): Vector store or database for long-term storage

```
session_id: str,
  memory_type: str = "short_term",
) -> MemoryEntry:
  """Create a new memory entry with metadata"""
  metadata = MemoryMetadata(
    timestamp=time.time(),
     role=role,
    agent_name=agent_name,
    session_id=session_id,
     memory_type=memory_type,
    token_count=self.tokenizer.count_tokens(content),
  )
  return MemoryEntry(content=content, metadata=metadata)
def add_memory(
  self,
  content: str,
  role: str,
  agent_name: str,
  session_id: str,
  is_system: bool = False,
) -> None:
  """Add a new memory entry to appropriate storage"""
  entry = self.create_memory_entry(
    content=content,
     role=role,
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agent_name=agent_name,
    session_id=session_id,
    memory_type="system" if is_system else "short_term",
  )
  if is_system:
    self.system_messages.append(entry)
  else:
    self.short_term_memory.append(entry)
  # Check if archiving is needed
  if self.should_archive():
    self.archive_old_memories()
  self.total_tokens_processed += entry.metadata.token_count
def get_current_token_count(self) -> int:
  """Get total tokens in short-term memory"""
  return sum(
    entry.metadata.token_count
    for entry in self.short_term_memory
  )
def get_system_messages_token_count(self) -> int:
  """Get total tokens in system messages"""
  return sum(
```

```
entry.metadata.token_count
    for entry in self.system_messages
  )
def should_archive(self) -> bool:
  """Check if archiving is needed based on configuration"""
  if not self.config.auto_archive:
     return False
  current_usage = (
    self.get_current_token_count()
    / self.config.max_short_term_tokens
  )
  return current_usage >= self.config.archive_threshold
def archive_old_memories(self) -> None:
  """Move older memories to long-term storage"""
  if not self.long_term_memory:
    logger.warning(
       "No long-term memory storage configured for archiving"
    )
    return
  while self.should_archive():
    # Get oldest non-system message
    if not self.short_term_memory:
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break
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```
oldest_entry = self.short_term_memory.pop(0)
    # Store in long-term memory
    self.store_in_long_term_memory(oldest_entry)
    self.archived_entries_count += 1
def store_in_long_term_memory(self, entry: MemoryEntry) -> None:
  """Store a memory entry in long-term memory"""
  if self.long_term_memory is None:
    logger.warning(
       "Attempted to store in non-existent long-term memory"
     )
     return
  try:
    self.long_term_memory.add(str(entry.model_dump()))
  except Exception as e:
    logger.error(f"Error storing in long-term memory: {e}")
    # Re-add to short-term if storage fails
    self.short_term_memory.insert(0, entry)
def get_relevant_context(
  self, query: str, max_tokens: Optional[int] = None
) -> str:
```

Get relevant context from both memory types

```
Args:
  query (str): Query to match against memories
  max_tokens (Optional[int]): Maximum tokens to return
Returns:
  str: Combined relevant context
contexts = []
# Add system messages first
for entry in self.system_messages:
  contexts.append(entry.content)
# Add short-term memory
for entry in reversed(self.short_term_memory):
  contexts.append(entry.content)
# Query long-term memory if available
if self.long_term_memory is not None:
  long_term_context = self.long_term_memory.query(query)
  if long_term_context:
    contexts.append(str(long_term_context))
```

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# Combine and truncate if needed
  combined = "\n".join(contexts)
  if max_tokens:
     combined = self.truncate_to_token_limit(
       combined, max_tokens
    )
  return combined
def truncate_to_token_limit(
  self, text: str, max_tokens: int
) -> str:
  """Truncate text to fit within token limit"""
  current_tokens = self.tokenizer.count_tokens(text)
  if current_tokens <= max_tokens:
     return text
  # Truncate by splitting into sentences and rebuilding
  sentences = text.split(". ")
  result = []
  current\_count = 0
  for sentence in sentences:
     sentence_tokens = self.tokenizer.count_tokens(sentence)
     if current_count + sentence_tokens <= max_tokens:</pre>
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result.append(sentence)
       current_count += sentence_tokens
    else:
       break
  return ". ".join(result)
def clear_short_term_memory(
  self, preserve system: bool = True
) -> None:
  """Clear short-term memory with option to preserve system messages"""
  if not preserve_system:
    self.system_messages.clear()
  self.short_term_memory.clear()
  logger.info(
    "Cleared short-term memory"
    + " (preserved system messages)"
    if preserve_system
    else ""
  )
def get_memory_stats(self) -> Dict[str, Any]:
  """Get detailed memory statistics"""
  return {
     "short_term_messages": len(self.short_term_memory),
     "system_messages": len(self.system_messages),
```

```
"system_tokens": self.get_system_messages_token_count(),
     "max_tokens": self.config.max_short_term_tokens,
     "token_usage_percent": round(
       (
         self.get_current_token_count()
         / self.config.max_short_term_tokens
       )
       * 100,
       2,
    ),
     "has_long_term_memory": self.long_term_memory is not None,
     "archived_entries": self.archived_entries_count,
     "total_tokens_processed": self.total_tokens_processed,
  }
def save_memory_snapshot(self, file_path: str) -> None:
  """Save current memory state to file"""
  try:
    data = {
       "timestamp": datetime.now().isoformat(),
       "config": self.config.model_dump(),
       "system_messages": [
         entry.model_dump()
         for entry in self.system_messages
       ],
```

"current_tokens": self.get_current_token_count(),

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"short_term_memory": [
          entry.model_dump()
          for entry in self.short_term_memory
       ],
       "stats": self.get_memory_stats(),
    }
     with open(file_path, "w") as f:
       if file_path.endswith(".yaml"):
          yaml.dump(data, f)
       else:
          json.dump(data, f, indent=2)
     logger.info(f"Saved memory snapshot to {file_path}")
  except Exception as e:
     logger.error(f"Error saving memory snapshot: {e}")
     raise
def load_memory_snapshot(self, file_path: str) -> None:
  """Load memory state from file"""
  try:
     with open(file_path, "r") as f:
       if file_path.endswith(".yaml"):
          data = yaml.safe_load(f)
       else:
```

```
data = json.load(f)
    self.config = MemoryConfig(**data["config"])
     self.system_messages = [
       MemoryEntry(**entry)
       for entry in data["system_messages"]
    ]
    self.short_term_memory = [
       MemoryEntry(**entry)
       for entry in data["short_term_memory"]
    ]
     logger.info(f"Loaded memory snapshot from {file_path}")
  except Exception as e:
     logger.error(f"Error loading memory snapshot: {e}")
     raise
def search_memories(
  self, query: str, memory_type: str = "all"
) -> List[MemoryEntry]:
  Search through memories of specified type
  Args:
    query (str): Search query
```

```
memory_type (str): Type of memories to search ("short_term", "system", "long_term", or "all")
```

```
Returns:
  List[MemoryEntry]: Matching memory entries
results = []
if memory_type in ["short_term", "all"]:
  results.extend(
     [
       entry
       for entry in self.short_term_memory
       if query.lower() in entry.content.lower()
     ]
  )
if memory_type in ["system", "all"]:
  results.extend(
    [
       entry
       for entry in self.system_messages
       if query.lower() in entry.content.lower()
     ]
  )
```

if (

```
memory_type in ["long_term", "all"]
  and self.long_term_memory is not None
):
  long_term_results = self.long_term_memory.query(query)
  if long_term_results:
    # Convert long-term results to MemoryEntry format
    for result in long_term_results:
       content = str(result)
       metadata = MemoryMetadata(
         timestamp=time.time(),
         role="long_term",
         agent_name="system",
         session_id="long_term",
         memory_type="long_term",
         token_count=self.tokenizer.count_tokens(
            content
         ),
       )
       results.append(
         MemoryEntry(
            content=content, metadata=metadata
         )
       )
```

return results

```
def get_memory_by_timeframe(
  self, start_time: float, end_time: float
) -> List[MemoryEntry]:
  """Get memories within a specific timeframe"""
  return [
     entry
    for entry in self.short_term_memory
    if start_time <= entry.metadata.timestamp <= end_time
  ]
def export_memories(
  self, file_path: str, format: str = "json"
) -> None:
  """Export memories to file in specified format"""
  data = {
     "system_messages": [
       entry.model_dump() for entry in self.system_messages
    ],
     "short_term_memory": [
       entry.model_dump() for entry in self.short_term_memory
    ],
     "stats": self.get_memory_stats(),
  }
  with open(file_path, "w") as f:
     if format == "yaml":
```

```
yaml.dump(data, f)
else:
    json.dump(data, f, indent=2)
```