```
import uuid
from collections import Counter
from datetime import datetime
from typing import Any, List, Optional
from pydantic import BaseModel, Field
from swarms.structs.agent import Agent
from swarms.utils.loguru_logger import initialize_logger
from swarms.utils.auto_download_check_packages import (
  auto_check_and_download_package,
)
from swarms.structs.conversation import Conversation
logger = initialize_logger(log_folder="tree_swarm")
# Pydantic Models for Logging
class AgentLogInput(BaseModel):
  log_id: str = Field(
    default_factory=lambda: str(uuid.uuid4()), alias="id"
  )
  agent_name: str
  task: str
  timestamp: datetime = Field(default_factory=datetime.utcnow)
```

```
class AgentLogOutput(BaseModel):
  log_id: str = Field(
     default_factory=lambda: str(uuid.uuid4()), alias="id"
  )
  agent_name: str
  result: Any
  timestamp: datetime = Field(default_factory=datetime.utcnow)
class TreeLog(BaseModel):
  log_id: str = Field(
     default_factory=lambda: str(uuid.uuid4()), alias="id"
  )
  tree_name: str
  task: str
  selected_agent: str
  timestamp: datetime = Field(default_factory=datetime.utcnow)
  result: Any
def extract_keywords(prompt: str, top_n: int = 5) -> List[str]:
  ....
  A simplified keyword extraction function using basic word splitting instead of NLTK tokenization.
  ....
  words = prompt.lower().split()
```

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filtered_words = [word for word in words if word.isalnum()]
  word_counts = Counter(filtered_words)
  return [word for word, _ in word_counts.most_common(top_n)]
class TreeAgent(Agent):
  111111
  A specialized Agent class that contains information about the system prompt's
  locality and allows for dynamic chaining of agents in trees.
  ....
  def __init__(
     self,
     name: str = None,
     description: str = None,
     system_prompt: str = None,
     model_name: str = "gpt-4o",
     agent_name: Optional[str] = None,
     *args,
     **kwargs,
  ):
     agent_name = agent_name
     super().__init__(
       name=name,
       description=description,
       system_prompt=system_prompt,
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```
model_name=model_name,
  agent_name=agent_name,
  *args,
  **kwargs,
)
try:
  import sentence_transformers
except ImportError:
  auto_check_and_download_package(
    "sentence-transformers", package_manager="pip"
  )
  import sentence_transformers
self.sentence_transformers = sentence_transformers
# Pretrained model for embeddings
self.embedding_model = (
  sentence_transformers.SentenceTransformer(
    "all-MiniLM-L6-v2"
  )
)
self.system_prompt_embedding = self.embedding_model.encode(
  system_prompt, convert_to_tensor=True
)
```

```
self.relevant_keywords = extract_keywords(system_prompt)
  # Distance is now calculated based on similarity between agents' prompts
  self.distance = None # Will be dynamically calculated later
def calculate_distance(self, other_agent: "TreeAgent") -> float:
  ....
  Calculate the distance between this agent and another agent using embedding similarity.
  Args:
    other_agent (TreeAgent): Another agent in the tree.
  Returns:
    float: Distance score between 0 and 1, with 0 being close and 1 being far.
  ....
  similarity = self.sentence_transformers.util.pytorch_cos_sim(
     self.system_prompt_embedding,
    other_agent.system_prompt_embedding,
  ).item()
  distance = (
     1 - similarity
  ) # Closer agents have a smaller distance
  return distance
def run_task(
```

# Automatically extract keywords from system prompt

```
self, task: str, img: str = None, *args, **kwargs
) -> Any:
  input_log = AgentLogInput(
     agent_name=self.agent_name,
     task=task,
     timestamp=datetime.now(),
  )
  logger.info(f"Running task on {self.agent_name}: {task}")
  logger.debug(f"Input Log: {input_log.json()}")
  result = self.run(task=task, img=img, *args, **kwargs)
  output_log = AgentLogOutput(
     agent_name=self.agent_name,
     result=result,
    timestamp=datetime.now(),
  )
  logger.info(f"Task result from {self.agent_name}: {result}")
  logger.debug(f"Output Log: {output_log.json()}")
  return result
def is_relevant_for_task(
  self, task: str, threshold: float = 0.7
) -> bool:
  .....
```

Checks if the agent is relevant for the given task using both keyword matching and embedding similarity.

```
Args:
  task (str): The task to be executed.
  threshold (float): The cosine similarity threshold for embedding-based matching.
Returns:
  bool: True if the agent is relevant, False otherwise.
# Check if any of the relevant keywords are present in the task (case-insensitive)
keyword_match = any(
  keyword.lower() in task.lower()
  for keyword in self.relevant_keywords
)
# Perform embedding similarity match if keyword match is not found
if not keyword_match:
  task_embedding = self.embedding_model.encode(
    task, convert_to_tensor=True
  )
  similarity = (
    self.sentence_transformers.util.pytorch_cos_sim(
       self.system_prompt_embedding, task_embedding
    ).item()
  )
```

```
logger.info(
          f"Semantic similarity between task and {self.agent_name}: {similarity:.2f}"
       )
       return similarity >= threshold
     return True # Return True if keyword match is found
class Tree:
  def __init__(self, tree_name: str, agents: List[TreeAgent]):
     11 11 11
     Initializes a tree of agents.
     Args:
       tree_name (str): The name of the tree.
       agents (List[TreeAgent]): A list of agents in the tree.
     self.tree_name = tree_name
     self.agents = agents
     self.calculate_agent_distances()
  def calculate_agent_distances(self):
       Automatically calculate and assign distances between agents in the tree based on prompt
similarity.
     .....
```

```
logger.info(
    f"Calculating distances between agents in tree '{self.tree_name}'"
  )
  for i, agent in enumerate(self.agents):
     if i > 0:
       agent.distance = agent.calculate_distance(
          self.agents[i - 1]
       )
     else:
       agent.distance = 0 # First agent is closest
  # Sort agents by distance after calculation
  self.agents.sort(key=lambda agent: agent.distance)
def find_relevant_agent(self, task: str) -> Optional[TreeAgent]:
  ....
  Finds the most relevant agent in the tree for the given task based on its system prompt.
  Uses both keyword and semantic similarity matching.
  Args:
     task (str): The task or query for which we need to find a relevant agent.
  Returns:
     Optional[TreeAgent]: The most relevant agent, or None if no match found.
  logger.info(
```

```
f"Searching relevant agent in tree '{self.tree_name}' for task: {task}"
  )
  for agent in self.agents:
    if agent.is_relevant_for_task(task):
       return agent
  logger.warning(
    f"No relevant agent found in tree '{self.tree_name}' for task: {task}"
  )
  return None
def log_tree_execution(
  self, task: str, selected_agent: TreeAgent, result: Any
) -> None:
  11 11 11
  Logs the execution details of a tree, including selected agent and result.
  ....
  tree_log = TreeLog(
     tree_name=self.tree_name,
     task=task,
     selected_agent=selected_agent.agent_name,
     timestamp=datetime.now(),
     result=result,
  )
  logger.info(
    f"Tree '{self.tree_name}' executed task with agent '{selected_agent.agent_name}'"
  )
```

```
logger.debug(f"Tree Log: {tree_log.json()}")
```

```
class ForestSwarm:
  def __init__(
     self,
     name: str = "default-forest-swarm",
     description: str = "Standard forest swarm",
     trees: List[Tree] = [],
     shared_memory: Any = None,
     rules: str = None,
     *args,
     **kwargs,
  ):
     Initializes the structure with multiple trees of agents.
     Args:
       trees (List[Tree]): A list of trees in the structure.
     """
     self.name = name
     self.description = description
     self.trees = trees
     self.shared_memory = shared_memory
     self.save_file_path = f"forest_swarm_{uuid.uuid4().hex}.json"
     self.conversation = Conversation(
```

```
time_enabled=True,
     auto_save=True,
     save_filepath=self.save_file_path,
     rules=rules,
  )
def find_relevant_tree(self, task: str) -> Optional[Tree]:
  ....
  Finds the most relevant tree based on the given task.
  Args:
     task (str): The task or query for which we need to find a relevant tree.
  Returns:
     Optional[Tree]: The most relevant tree, or None if no match found.
  logger.info(
     f"Searching for the most relevant tree for task: {task}"
  )
  for tree in self.trees:
     if tree.find_relevant_agent(task):
       return tree
  logger.warning(f"No relevant tree found for task: {task}")
  return None
def run(self, task: str, img: str = None, *args, **kwargs) -> Any:
```

Executes the given task by finding the most relevant tree and agent within that tree.

```
Args:
  task (str): The task or query to be executed.
Returns:
  Any: The result of the task after it has been processed by the agents.
try:
  logger.info(
    f"Running task across MultiAgentTreeStructure: {task}"
  )
  relevant_tree = self.find_relevant_tree(task)
  if relevant_tree:
     agent = relevant_tree.find_relevant_agent(task)
     if agent:
       result = agent.run_task(
          task, img=img, *args, **kwargs
       )
       relevant_tree.log_tree_execution(
          task, agent, result
       return result
  else:
     logger.error(
```

```
"Task could not be completed: No relevant agent or tree found."
         )
         return "No relevant agent found to handle this task."
     except Exception as error:
       logger.error(
         f"Error detected in the ForestSwarm, check your inputs and try again;) {error}"
       )
## Example Usage:
# # Create agents with varying system prompts and dynamically generated distances/keywords
# agents_tree1 = [
    TreeAgent(
      system_prompt="Stock Analysis Agent",
      agent_name="Stock Analysis Agent",
    ),
    TreeAgent(
      system prompt="Financial Planning Agent",
      agent_name="Financial Planning Agent",
    ),
    TreeAgent(
      agent_name="Retirement Strategy Agent",
      system_prompt="Retirement Strategy Agent",
    ),
```

#

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#

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#

#

#]

```
# agents_tree2 = [
#
    TreeAgent(
      system_prompt="Tax Filing Agent",
#
#
      agent_name="Tax Filing Agent",
#
    ),
    TreeAgent(
#
      system_prompt="Investment Strategy Agent",
#
#
      agent_name="Investment Strategy Agent",
#
    ),
#
    TreeAgent(
#
      system_prompt="ROTH IRA Agent", agent_name="ROTH IRA Agent"
#
    ),
#]
## Create trees
# tree1 = Tree(tree_name="Financial Tree", agents=agents_tree1)
# tree2 = Tree(tree_name="Investment Tree", agents=agents_tree2)
## Create the ForestSwarm
# multi_agent_structure = ForestSwarm(trees=[tree1, tree2])
##Run a task
# task = "Our company is incorporated in delaware, how do we do our taxes for free?"
# output = multi_agent_structure.run(task)
# print(output)
```