

GroupChat Swarm Documentation

A production-grade multi-agent system enabling sophisticated group conversations between AI agents with customizable speaking patterns, parallel processing capabilities, and comprehensive conversation tracking.

Advanced Configuration

Agent Parameters

| Parameter | Type | Default | Description |
|---------------------|------|----------|---------------------------------|
| agent_name | str | Required | Unique identifier for the agent |
| system_prompt | str | Required | Role and behavior instructions |
| llm | Any | Required | Language model instance |
| max_loops | int | 1 | Maximum conversation turns |
| autosave | bool | False | Enable conversation saving |
| dashboard | bool | False | Enable monitoring dashboard |
| verbose | bool | True | Enable detailed logging |
| dynamic_temperature | bool | True | Enable dynamic temperature |
| retry_attempts | int | 1 | Failed request retry count |
| context_length | int | 200000 | Maximum context window |
| output_type | str | "string" | Response format type |
| streaming_on | bool | False | Enable streaming responses |

GroupChat Parameters

| Parameter | Type | Default | Description |
|-------------|-------------|-------------|----------------------------|
| name | str | "GroupChat" | Chat group identifier |
| description | str | "" | Purpose description |
| agents | List[Agent] | [] | Participating agents |
| speaker_fn | Callable | round_robin | Speaker selection function |
| max_loops | int | 10 | Maximum conversation turns |

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Installation

```
```bash
```

```
pip3 install swarms swarm-models loguru
```

## ## Core Concepts

The GroupChat system consists of several key components:

1. **Agents**: Individual AI agents with specialized knowledge and roles
2. **Speaker Functions**: Control mechanisms for conversation flow
3. **Chat History**: Structured conversation tracking
4. **Response Models**: Pydantic models for data validation

## ## Basic Usage

```
```python
```

```
import os
```

```
from dotenv import load_dotenv
```

```
from swarm_models import OpenAIChat
```

```
from swarms import Agent, GroupChat, expertise_based
```

```
if __name__ == "__main__":
```

```
    load_dotenv()
```

```
    # Get the OpenAI API key from the environment variable
```

```
api_key = os.getenv("OPENAI_API_KEY")
```

```
# Create an instance of the OpenAIChat class
```

```
model = OpenAIChat(  
    openai_api_key=api_key,  
    model_name="gpt-4o-mini",  
    temperature=0.1,  
)
```

```
# Example agents
```

```
agent1 = Agent(  
    agent_name="Financial-Analysis-Agent",  
    system_prompt="You are a financial analyst specializing in investment strategies.",  
    llm=model,  
    max_loops=1,  
    autosave=False,  
    dashboard=False,  
    verbose=True,  
    dynamic_temperature_enabled=True,  
    user_name="swarms_corp",  
    retry_attempts=1,  
    context_length=200000,  
    output_type="string",  
    streaming_on=False,  
)
```

```
agent2 = Agent(  
    agent_name="Tax-Adviser-Agent",  
    system_prompt="You are a tax adviser who provides clear and concise guidance on tax-related  
queries.",  
    llm=model,  
    max_loops=1,  
    autosave=False,  
    dashboard=False,  
    verbose=True,  
    dynamic_temperature_enabled=True,  
    user_name="swarms_corp",  
    retry_attempts=1,  
    context_length=200000,  
    output_type="string",  
    streaming_on=False,  
)
```

```
agents = [agent1, agent2]
```

```
chat = GroupChat(  
    name="Investment Advisory",  
    description="Financial and tax analysis group",  
    agents=agents,  
    speaker_fn=expertise_based,  
)
```

```

history = chat.run(
    "How to optimize tax strategy for investments?"
)

print(history.model_dump_json(indent=2))

...

```

Speaker Functions

Built-in Functions

```
``python
```

```
def round_robin(history: List[str], agent: Agent) -> bool:
```

```
    """
```

```
    Enables agents to speak in turns.
```

```
    Returns True for each agent in sequence.
```

```
    """
```

```
    return True
```

```
def expertise_based(history: List[str], agent: Agent) -> bool:
```

```
    """
```

```
    Enables agents to speak based on their expertise.
```

```
    Returns True if agent's role matches conversation context.
```

```
    """
```

```
    return agent.system_prompt.lower() in history[-1].lower() if history else True
```

```
def random_selection(history: List[str], agent: Agent) -> bool:
```

```
    """
```

```
    Randomly selects speaking agents.
```

```
    Returns True/False with 50% probability.
```

```
    """
```

```
    import random
```

```
    return random.choice([True, False])
```

```
def most_recent(history: List[str], agent: Agent) -> bool:
```

```
    """
```

```
    Enables agents to respond to their mentions.
```

```
    Returns True if agent was last speaker.
```

```
    """
```

```
    return agent.agent_name == history[-1].split(":")[0].strip() if history else True
```

```
...
```

Custom Speaker Function Example

```
```python
```

```
def custom_speaker(history: List[str], agent: Agent) -> bool:
```

```
 """
```

```
 Custom speaker function with complex logic.
```

Args:

history: Previous conversation messages

agent: Current agent being evaluated

Returns:

bool: Whether agent should speak

"""

# No history - let everyone speak

if not history:

return True

last\_message = history[-1].lower()

# Check for agent expertise keywords

expertise\_relevant = any(

keyword in last\_message

for keyword in agent.expertise\_keywords

)

# Check for direct mentions

mentioned = agent.agent\_name.lower() in last\_message

# Check if agent hasn't spoken recently

not\_recent\_speaker = not any(

agent.agent\_name in msg

for msg in history[-3:]

)



return expertise\_relevant or mentioned or not\_recent\_speaker

# Usage

```
chat = GroupChat(
 agents=[agent1, agent2],
 speaker_fn=custom_speaker
)
...

```

## Response Models

### Complete Schema

```
```python

```

```
class AgentResponse(BaseModel):
    """Individual agent response in a conversation turn"""
    agent_name: str
    role: str
    message: str
    timestamp: datetime = Field(default_factory=datetime.now)
    turn_number: int
    preceding_context: List[str] = Field(default_factory=list)

```

```
class ChatTurn(BaseModel):
    """Single turn in the conversation"""
    turn_number: int

```

responses: List[AgentResponse]

task: str

timestamp: datetime = Field(default_factory=datetime.now)

class ChatHistory(BaseModel):

"""Complete conversation history"""

turns: List[ChatTurn]

total_messages: int

name: str

description: str

start_time: datetime = Field(default_factory=datetime.now)

...

Advanced Examples

Multi-Agent Analysis Team

```
```python
```

```
Create specialized agents
```

```
data_analyst = Agent(
```

```
 agent_name="Data-Analyst",
```

```
 system_prompt="You analyze numerical data and patterns",
```

```
 llm=model
```

```
)
```

```
market_expert = Agent(
```

```
agent_name="Market-Expert",
system_prompt="You provide market insights and trends",
llm=model
)
```

```
strategy_advisor = Agent(
 agent_name="Strategy-Advisor",
 system_prompt="You formulate strategic recommendations",
 llm=model
)
```

# Create analysis team

```
analysis_team = GroupChat(
 name="Market Analysis Team",
 description="Comprehensive market analysis group",
 agents=[data_analyst, market_expert, strategy_advisor],
 speaker_fn=expertise_based,
 max_loops=15
)
```

# Run complex analysis

```
history = analysis_team.run("""
```

Analyze the current market conditions:

1. Identify key trends
2. Evaluate risks
3. Recommend investment strategy

```
""")
```

```
...
```

### ### Parallel Processing

```
```python
```

```
# Define multiple analysis tasks
```

```
tasks = [
```

```
    "Analyze tech sector trends",
```

```
    "Evaluate real estate market",
```

```
    "Review commodity prices",
```

```
    "Assess global economic indicators"
```

```
]
```

```
# Run tasks concurrently
```

```
histories = chat.concurrent_run(tasks)
```

```
# Process results
```

```
for task, history in zip(tasks, histories):
```

```
    print(f"\nAnalysis for: {task}")
```

```
    for turn in history.turns:
```

```
        for response in turn.responses:
```

```
            print(f"{response.agent_name}: {response.message}")
```

```
...
```

Best Practices

1. ****Agent Design****

- Give agents clear, specific roles
- Use detailed system prompts
- Set appropriate context lengths
- Enable retries for reliability

2. ****Speaker Functions****

- Match function to use case
- Consider conversation flow
- Handle edge cases
- Add appropriate logging

3. ****Error Handling****

- Use try-except blocks
- Log errors appropriately
- Implement retry logic
- Provide fallback responses

4. ****Performance****

- Use concurrent processing for multiple tasks
- Monitor context lengths
- Implement proper cleanup
- Cache responses when appropriate

API Reference

GroupChat Methods

| Method | Description | Arguments | Returns |
|---------------------|-------------------------------|------------------|-------------------|
| run | Run single conversation | task: str | ChatHistory |
| batched_run | Run multiple sequential tasks | tasks: List[str] | List[ChatHistory] |
| concurrent_run | Run multiple parallel tasks | tasks: List[str] | List[ChatHistory] |
| get_recent_messages | Get recent messages | n: int = 3 | List[str] |

Agent Methods

| Method | Description | Returns |
|-------------------|---------------------------|---------|
| run | Process single task | str |
| generate_response | Generate LLM response | str |
| save_context | Save conversation context | None |