

# Theory Development

EES 4760/5760

Agent-Based & Individual-Based Computational Modeling

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Live web page: [https://ees4760.jgilligan.org/Slides/Class\\_21](https://ees4760.jgilligan.org/Slides/Class_21)

PDF: [https://ees4760.jgilligan.org/Slides/Class\\_21/EES\\_4760\\_5760\\_Class\\_21\\_Slides.pdf](https://ees4760.jgilligan.org/Slides/Class_21/EES_4760_5760_Class_21_Slides.pdf)

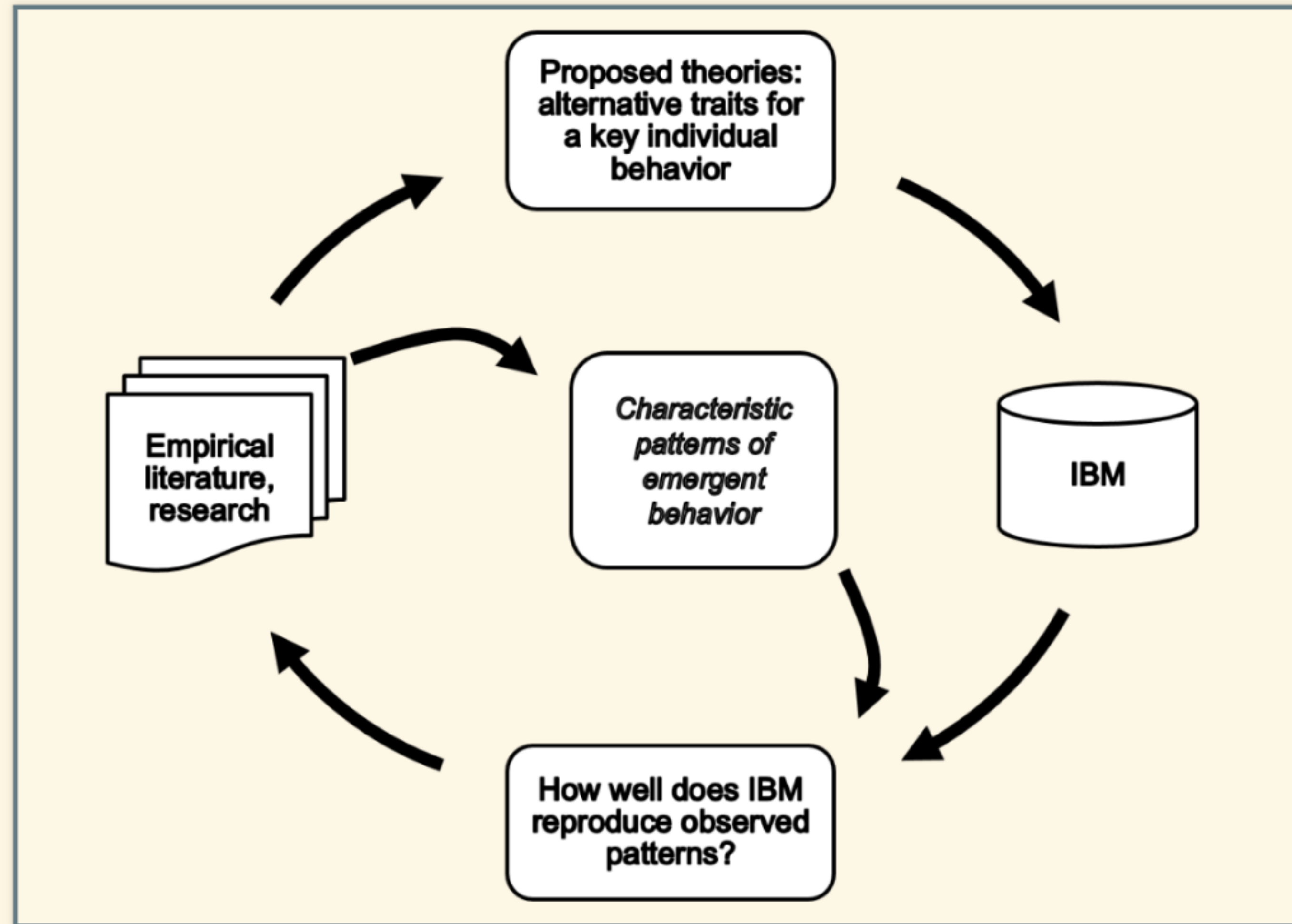
# Downloads

- Download page: [https://ees4760.jgilligan.org/downloads/hoopoe\\_class\\_21/](https://ees4760.jgilligan.org/downloads/hoopoe_class_21/)
- Zip file [https://ees4760.jgilligan.org/models/class\\_21/class\\_21\\_models.zip](https://ees4760.jgilligan.org/models/class_21/class_21_models.zip) containing Wood Hoopoe breeding model:
  - NetLogo model:  
[https://ees4760.jgilligan.org/models/class\\_21/wood\\_hoopoe\\_class\\_21.nlogo](https://ees4760.jgilligan.org/models/class_21/wood_hoopoe_class_21.nlogo)
  - ODD: [https://ees4760.jgilligan.org/models/class\\_21/wood\\_hoopoe\\_odd.pdf](https://ees4760.jgilligan.org/models/class_21/wood_hoopoe_odd.pdf)
  - NetLogo model with alternative strategies:  
[https://ees4760.jgilligan.org/models/class\\_21/wood\\_hoopoe\\_strategies.nlogo](https://ees4760.jgilligan.org/models/class_21/wood_hoopoe_strategies.nlogo)

# Models as a Virtual Laboratory

- How to use models to run experiments?
- Strong inference (John Platt)
- Identify traits (individual behaviors) that give rise to multiple macroscopic patterns
  1. Identify alternative traits (hypotheses)
  2. Implement traits in ABM
  3. Test and compare alternatives:
    - How well did model reproduce observed patterns?
    - Falsify traits that did not reproduce patterns
  4. Repeat cycle as needed. Revise behavior traits, look for additional patterns, etc.

# Pattern-Oriented Modeling Cycle





# Example: Trader intelligence

## Continuous Double Auction

1. Traders establish buying and selling prices
  - If someone offers a price  $\geq$  selling price, trader sells.
  - If someone offers to sell for  $\leq$  buying price, trader buys
2. Match traders:
  - If traders  $i$  and  $j$  have  $P_{i,\text{sell}} \leq P_{j,\text{buy}}$ , then transaction occurs.

## Zero-intelligence agent

- Agent sets random buying and selling price
- If  $P_{i,\text{buy}} > P_{i,\text{sell}}$ , then trader  $i$  will lose money.

## Minimal-intelligence agent

- Random buying and selling price with constraint:  $P_{i,\text{buy}} < P_{i,\text{sell}}$ .

# Results

- Minimal-intelligence agent was better than zero-intelligence
  - Zero-intelligence produced wild price fluctuations
  - Minimal-intelligence reproduced observed pattern of rapid price convergence
  - Minimal-intelligence also reproduced observed effects of price-ceiling.
- But simple models had limits:
  - Observed volatility of lower-end prices was not reproduced by models
  - As experimental markets got more complicated, human traders did worse, but models did ***much*** worse.

## Lessons

*Using zero-intelligence as a baseline, the researcher can ask:  
what is the minimal additional structure or restrictions on agent  
behavior that are necessary to achieve a certain goal.*



# Example: Harvesting Common Resource

- Experimental subjects move avatars on screen to harvest tokens (like simple video game)
- Players compete to get most tokens
- Tokens grow back at some rate
- Patterns:
  1. Number of tokens on screen over time
  2. Inequality between players
  3. # tokens collected in first four minutes
  4. Number of straight-line moves

# Theory development

1. Näive model: (random) Moves randomly
2. Näive model: (greedy) Always goes to nearest token
3. Clever model:
  - Prefers nearby tokens
  - Prefers clusters of tokens
  - Prefers tokens straight ahead
  - Avoids tokens close to other players
- Näive models do not match any of the four patterns.
- Ran clever model 100 times for each of 65,536 different combinations of parameters that characterize preferences.
  - Only 37 combinations of parameters matched all four patterns in data.
  - Patterns 2 and 3 are seen for most parameter values
  - Patterns 1 and 4 seen less frequently
  - Therefore:
    - Patterns 2 and 3 are built into the structure of the game.
    - Patterns 1 and 4 may give insight into human behavior.



# Example: Woodhoopoe



[https://ees4760.jgilligan.org/models/class\\_21/wood\\_hoopoe\\_class\\_21.nlogo](https://ees4760.jgilligan.org/models/class_21/wood_hoopoe_class_21.nlogo)

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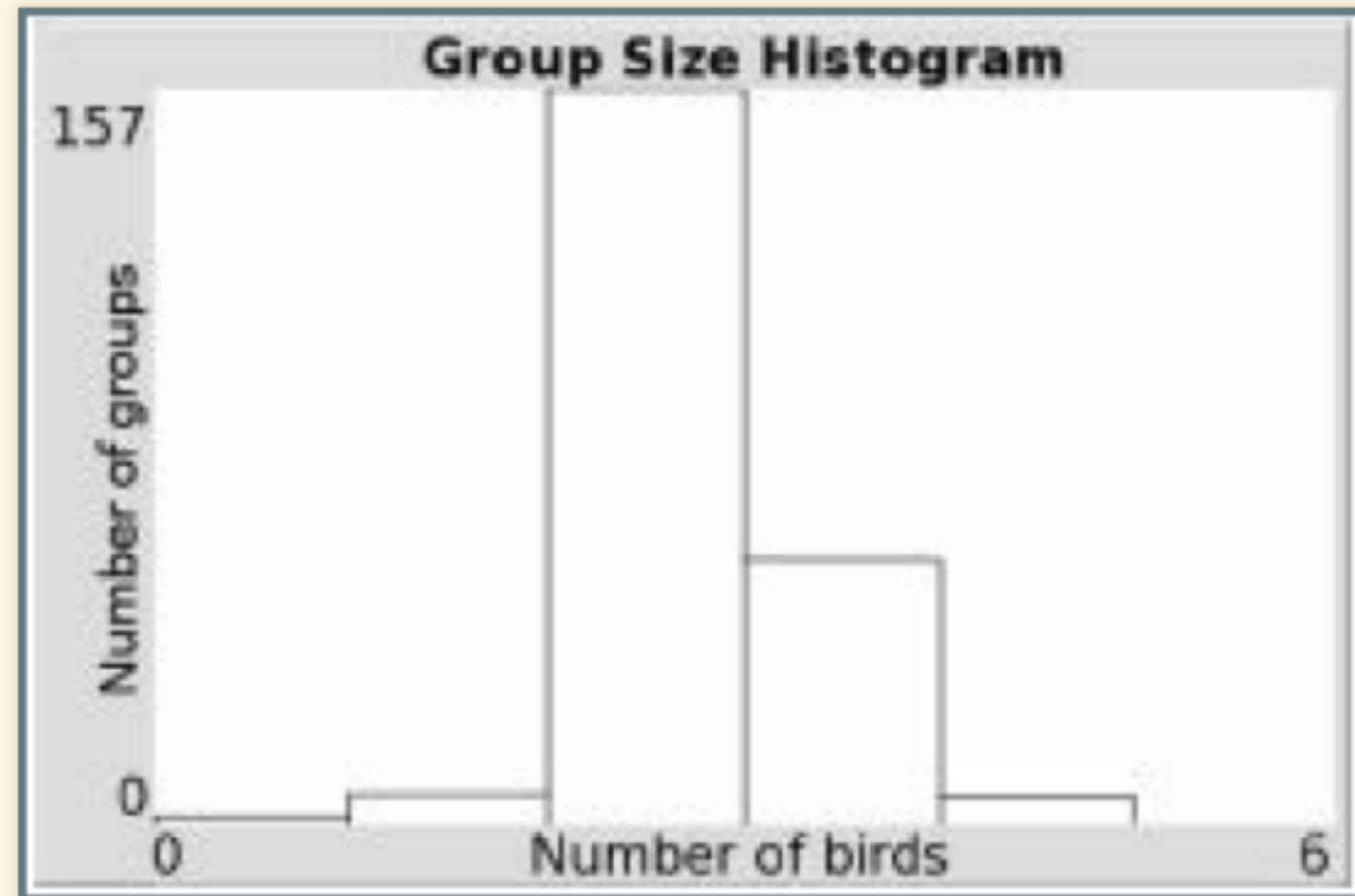
# Observed Behaviors

- Groups occupy spatial territories
- One **alpha** of each sex in a territory
- Only alpha couple reproduces
- If alpha dies, oldest subordinate of that sex becomes alpha
- **Scouting forays**
  - Subordinate adult leaves territory
  - If it finds territory without alpha, it stays, becomes alpha
  - Otherwise, returns home
  - Risk of predation (death) is high on scouting forays
- Alpha couple breeds once a year, in December



# Observed Patterns

1. Characteristic group size distribution (adults)



2. Average age of birds on scouting forays is younger than average age of all subordinates.
3. Scouting forays most common April–October

# Modeling Woodhoopoe

- Start simple:
  - One-dimensional world
  - One tick = one month
  - Every tick, bird has 1% chance of dying (0.99 probability to survive)
  - Scouting forays have 20% chance of death (0.80 probability to survive)
  - Adult subordinates go scouting at random (50% probability each tick)
- Does model reproduce patterns?

# Developing Alternative Strategies

[https://ees4760.jgilligan.org/models/class\\_21/wood\\_hoopoe\\_strategies.nlogo](https://ees4760.jgilligan.org/models/class_21/wood_hoopoe_strategies.nlogo)