

Computational Social Science

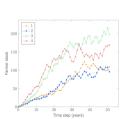


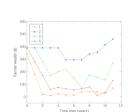


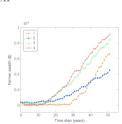
851-0101-86 S Complex Social Systems: Modeling Agents, Learning, and Games

CropWar: Agent-based simulation of agricultural interactions

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CropWar aims at modelling interactions between farmers growing crops in a geographic area with a finite amount of water ressources and selling them on a market to feed a population.



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v1.1 basic version: growing, selling, stocking



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- ▶ v2.2 reinforcement learning: identifying the optimal strategy against Trader agents

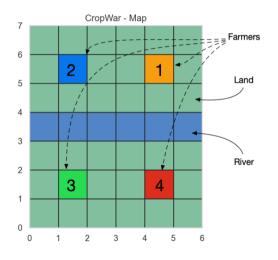
Basic version: selling and stocking



v1.1: Basic version

Basic version: selling and stocking

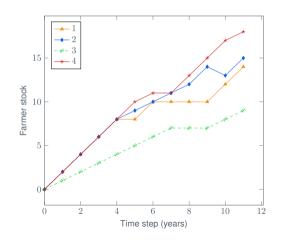




In v1.1 farmers only have access to one cell on which they can choose to grow one of two crops (A or B)

Basic version: selling and stocking

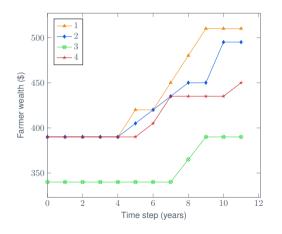




This plot shows the evolution of stock as a function of time:
Farmers 1,2 and 4 grow crop A
Farmer 3 grows crop B
At each time step, they chose to stock or sell their yield

Basic version: selling and stocking





This plot shows the corresponding evolution of the farmer's wealth as a function of time. Selling corresponds to an increase in wealth, stocking to a constant value.

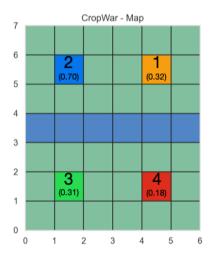
The Map class: expansion



v1.2: Spatial expansion

The Map class: expansion

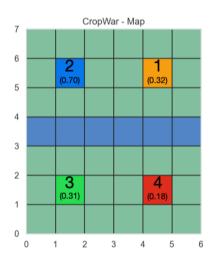




The buy_threshold (in brackets) indicates the tendancy of farmers to expand their farms by buying neighbouring land in order to grow more crops

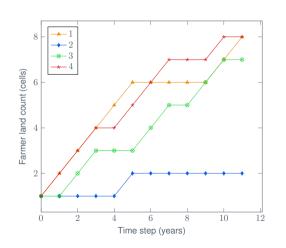
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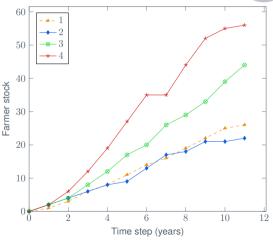




The Map class: expansion

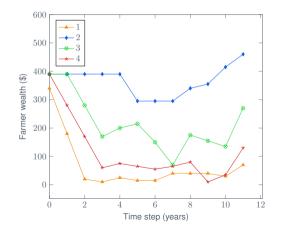






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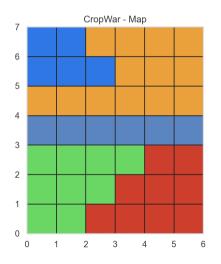




Farmer 2 seems to remain wealthier than other farmers over 10 time steps, due to not spending money to expand

The Map class: expansion

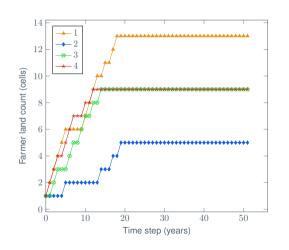


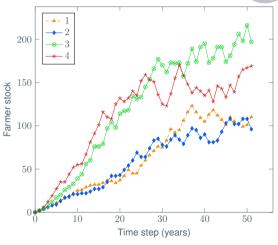


If we extend the simulation to 50 time steps, all available land is bought by farmers

The Map class: expansion

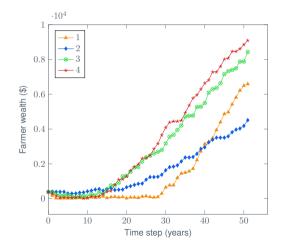






The Map class: expansion





On the long run, farmers with more land become wealthier, although the type of crop also seems to have an importance

The Market model: dynamic pricing



v1.3: Dynamic pricing of crops

The Market model: dynamic pricing



In v1.3 we implement a Market class, which updates the price of crops based on supply and demand.

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- Harvesting period: agents harvest according to their crop choice and add the harvest yield to the stock.
- 2. Interaction period: market interaction takes place.
- 3. Strategy period: the Agent's personality influences its decision for the next time step.

The Market model: dynamic pricing



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The Market model: dynamic pricing



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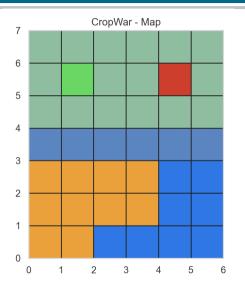
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- Stationary agents remain on their initial cell
- Expanding agents may buy neighbouring land

The Market model: dynamic pricing

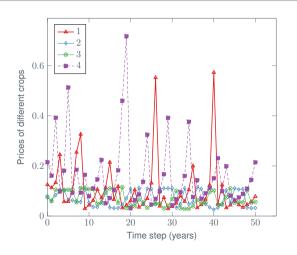


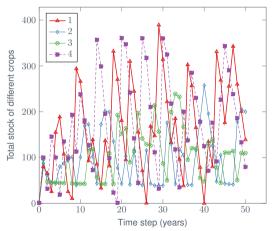


Simulation with 2 stationary Introverts and 2 expanding Traders $\,$

The Market model: dynamic pricing

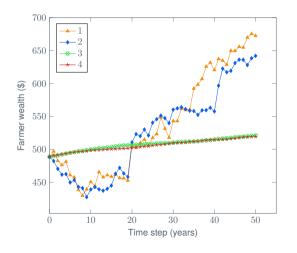






The Market model: dynamic pricing





Farmer wealth as a function of time: expanding Traders agents take more risks, which seem to pay off on the long term

Reinforcement Learning versions



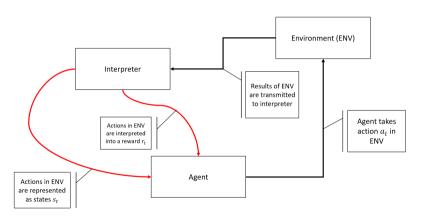
Reinforcement Learning



The preceding versions v1.1 - v1.3 defined the deterministic agents and the model of the environment in which actions take place.

In order to see unpredictable, emerging behaviour, the agents would need to adjust to evolutions of the market and learn.





Feedback loop in RL. An agent decides to do a certain action a_t based on his observation of the environment. This affects the environment, which yields a new state. The effect of the agents' action is then interpreted to update the strategy of the agent.



$$\underbrace{L(\theta)}_{\begin{subarray}{c} \begin{subarray}{c} \begin{subarra$$

Proximal Policy Optimization



$$\underbrace{L(\theta)}_{\substack{\text{Policy} \\ \text{loss} \\ \text{function}}} = \underbrace{\hat{\mathbf{E}}_t}_{\substack{\text{expectation} \\ \text{at iteration} \\ t}} \underbrace{\log(\pi_{\theta}(a_t|s_t))}_{\substack{\text{stochastic} \\ \text{policy} \\ \text{policy}}} \underbrace{\hat{A}_t}_{\substack{\text{estimator of} \\ \text{advantage of} \\ \text{current action}}} \tag{1}$$

where:



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 \blacktriangleright $\pi_{\theta}(a_t|s_t)$ is given as transition probabilities in the MC of taking action a_t in state s_t . It is a neural network, that suggests an action for a given state based on previous training experience.



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- \hat{A}_t is computed as a discounted reward. This neural net is updated with the experience (i.e. reward) that the agent collects in an environment.

Reinforcement Learning versions ML agent trained against introverts



v2.1: RL agent trained againt Introvert farmers



The reward function that was used for training is given by

$$\frac{1}{3}(\underbrace{b^2}_{\text{Fraction of Global Budget}} + \underbrace{r^2}_{\text{Current Ranking}} + \underbrace{s^2}_{\text{Global supply}}) \tag{2}$$

where r is 1 for first, 0.66 for second, 0.33 for third, and 0.0 for last agent.

Reinforcement Learning versions ML agent trained against introverts



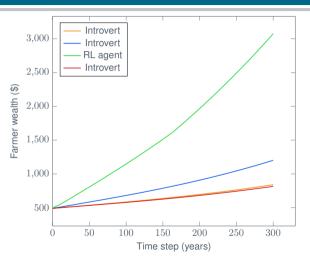


Figure: Budget evolution over 300 iterations

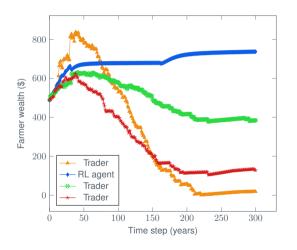
Reinforcement Learning versions ML agent trained against traders

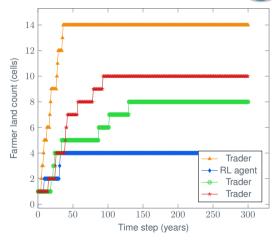
CropWar 28

v2.1: RL agent trained against Trader farmers

Reinforcement Learning versions ML agent trained against traders







Reinforcement Learning versions ML agent trained against traders



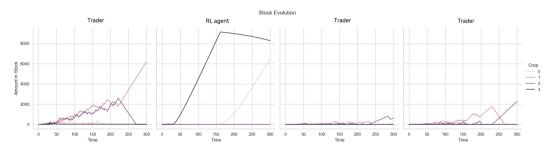


Figure: Stock evolution for the three Trader agents and the RL agent.

Crop War in perspective



Work Flow to improve the model



We organised the model to allow for an easy code review and potential improvement by other people:



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Crop War in perspective



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- ► The Python code can be edited to play around with different parameters and equations and a Documentation was created to allow for an easier code review.





Possible extensions include:

▶ More realistic spending of ressources: equipment investments, maintenance costs, etc...



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- ► Bigger and more realistic maps



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- Bigger and more realistic maps
- Weather events
- Ecological impact of different crops via pesticide use for example



CropWar

