

# Module

# axelrod\_model\_improvement\_groupdependentat

Simulations based on the Axelrod model and extensions to inspect the volatility of the features over time (AXELROD MODEL + Agreement threshold + two model variations based on the Social identity approach) The Axelrod model is used to predict the number of changes per feature in comparison to the datasets and is used to compare different model variations and their performance.

Input: Real data

–take the number of change per step and not the distance function between the feature vectors as a break criteria vector-version of axelrod –

Description: The Axelrod model depicts convergence and diversity on a macro level, driven by local agent-based interaction mechanisms. The agreement-threshold model (MacCarron et al. 2020a), an extension of the Axelrod model, acts as a multi-dimensional opinion dynamics model. We extend these agent-based models by explicit aspects of the social identity approach to recover real-world dynamics better and to assess the prediction performance of data simulations. We newly introduce mechanisms on in- and out-group interaction.

Model variations: Parameter: Group-dependent preference - Interaction within a group takes place without the limitation of the agreement threshold - Interaction between groups involves an agreement threshold - Group parameter needed

Parameter: In-group preference - Integrates an interaction preference towards in-group members - Reduced interaction probability with an out-group members - Inter-group interaction still possible, but unlikely - Group parameter needed

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## Functions

```
def count_cluster(f)
```

```
    count the number of clusters (agents with identical vectors) #no used#
```

### Args

```
f : np.array  
    cultural vectors of the agents
```

## Returns

`list`

mean cluster size, number of clusters

```
def feature_distribution(n_features=8, features=None, group_array=None)
```

get feature distribution for each group

## Args

**n\_features** : `int`, optional

*description*. Defaults to 8.

**features** : `_type_`, optional

*description*. Defaults to None.

**group\_array** : `_type_`, optional

*description*. Defaults to None.

## Raises

`ValueError`

("group and features do not have the same length")

## Returns

`_type_`

(dict) with attitude distributions

```
def number_of_changes(dataset, precise_difference=True)
```

define number of changes and give back a dictionary

## Args

**dataset** : `string`

filename

**precise\_difference** : `bool`, optional

precise difference (forth and back are two changes) Defaults to True.

## Returns

`_type_`

(dict) differences, (dict) used data, ID\_party\_aff, diff\_array

```
def run_main(selected_dataset='BJSP', at_value=[1], topo_value=[0], rounds=2000,  
            only_same_group_interaction=False, group_dependent_at=False)
```

main function

## Args

**selected\_dataset** : str , optional  
determine dataset. Defaults to 'BJSP'.

**at\_value** : list , optional  
agreement threshold. Defaults to [1].

**topo\_value** : list , optional  
topology of the model. Defaults to [0].

**rounds** : int , optional  
max number of simulation runs. Defaults to 2000.

**only\_same\_group\_interaction** : bool , optional  
model variation 1. Defaults to False.

**group\_dependent\_at** : bool , optional  
model variation 2. Defaults to False.

```
def run_model(grid, n_features, t_max, objective_changes_list, topo, at, connection,  
            timepoint_to_measure, data_array, group_array, only_same_group_interaction,  
            group_dependent_at, q)
```

callable run function for multiprocessing

## Args

**grid** : int  
size of grid

**n\_features** : int  
number of features

**t\_max** : int  
max running time

**objective\_changes\_list** : list  
maximum number of changes which are simulated by the mode

**topo** : int  
underlying topology of the model

**at** : int  
agreement threshold; 0 = no agreement threshold

**connection** : int  
number of links between agents (for example to generate AB network)

**timepoint\_to\_measure** : int  
How often do we measure the state of the model

**data\_array** : np.array  
position of each agent from the data

**group\_array** : np.array  
group identifies for every agent

**only\_same\_group\_interaction** : bool  
model variation; interaction highly limited to group members

**group\_dependent\_at** : bool  
model variation; at for out-group members and no at for in-group members

**q** : queue  
multiprocessing, get back the results (return statement)

## Classes

```
class AxelrodModelNumpy (Grid_x=10, Grid_y=10, n_features=8, t_max=100,  
                        objective_changes_list=[1000], topo=0, at=0, m_AB=2, memory=False,  
                        timepoint_to_measure=10000, data_array=None, group_array=None,  
                        only_same_group_interaction=False, group_dependent_at=False)
```

## Methods

```
def run_model(self, t_max)
```

Run method of the Axelrod model, is dependent on the number of changes from the data and will stop when the maximum number of possible changes is reached. We only set a limit to secure that it terminates but usually it should not be reached. The model runs and gives the results for :param t\_max: number of events :return: set of measurements → list(number of changes per feature, length of run, number of overall changes, biggest cluster size)

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