

---

# **ChannelAttribution**

***Release 2.2.2***

**Davide Altomare, David Loris**

**Nov 06, 2025**



**CONTENTS:**

<b>1</b>	<b>Indices and tables</b>	<b>1</b>
<b>2</b>	<b>ChannelAttribution</b>	<b>3</b>
	<b>Python Module Index</b>	<b>11</b>
	<b>Index</b>	<b>13</b>



## INDICES AND TABLES

- `genindex`
- `modindex`
- `search`



## CHANNELATTRIBUTION

**Markov Model for Online Multi-Channel Attribution** Advertisers use a variety of online marketing channels to reach consumers and they want to know the degree each channel contributes to their marketing success. This is called online multichannel attribution problem. In many cases, advertisers approach this problem through some simple heuristics methods that do not take into account any customer interactions and often tend to underestimate the importance of small channels in marketing contribution. This package provides a function that approaches the attribution problem in a probabilistic way. It uses a k-order Markov representation to identify structural correlations in the customer journey data. This would allow advertisers to give a more reliable assessment of the marketing contribution of each channel. The approach basically follows the one presented in Eva Anderl, Ingo Becker, Florian v. Wangenheim, Jan H. Schumann (2014). Differently from them, we solved the estimation process using stochastic simulations. In this way it is also possible to take into account conversion values and their variability in the computation of the channel importance. The package also contains a function that estimates three heuristic models (first-touch, last-touch and linear-touch approach) for the same problem.

```
ChannelAttribution.auto_markov_model(Data, var_path, var_conv, var_null, var_value=None,
                                     max_order=10, roc_npt=100, plot=False, nsim_start=100000.0,
                                     max_step=None, out_more=False, sep='>', ncore=1, nfold=10,
                                     seed=0, conv_par=0.05, rate_step_sim=1.5, verbose=True,
                                     flg_pro=True)
```

### Parameters

- **Data** (*DataFrame*) – customer journeys.
- **var\_path** (*string*) – column of Data containing paths.
- **var\_conv** (*string*) – column of Data containing total conversions for each path.
- **var\_null** (*string*) – column of Data containing total paths that do not lead to conversion.
- **var\_value** (*string, optional, default None*) – column of Data containing revenue for each path
- **max\_order** (*int, default 10*) – maximum Markov Model order to be considered.
- **roc\_npt** (*int, default 100*) – number of points to be used for the approximation of roc curve.
- **plot** (*bool, default True*) – if True, a plot with penalized auc with respect to order will be displayed.
- **nsim\_start** (*int, default 1e5*) – minimum number of simulations to be used in computation.
- **max\_step** (*int, default None*) – maximum number of steps for a single simulated path. if NULL, it is the maximum number of steps found into Data.

- **out\_more** (*bool*, *default False*) – if True, transition probabilities between channels and removal effects will be returned.
- **sep** (*string*, *default ">"*) – separator between the channels.
- **ncore** (*int*, *default 1*) – number of threads to be used in computation.
- **nfold** (*int*, *default 10*) – how many repetitions to be used to verify if convergence has been reached at each iteration.
- **seed** (*int*, *default 0*) – random seed. Giving this parameter the same value over different runs guarantees that results will not vary.
- **conv\_par** (*double*, *default 0.05*) – convergence parameter for the algorithm. The estimation process ends when the percentage of variation of the results over different repetitions is less than convergence parameter.
- **rate\_step\_sim** (*double*, *default 0*) – number of simulations used at each iteration is equal to the number of simulations used at previous iteration multiplied by rate\_step\_sim.
- **verbose** (*bool*, *default True*) – if True, additional information about process convergence will be shown.
- **flg\_pro** (*bool*, *default True*) – if True, ChannelAttribution Pro banner is printed.

#### Returns

##### result: Dataframe

(column) channel\_name : channel names (column) total\_conversions : conversions attributed to each channel (column) total\_conversion\_value : revenues attributed to each channel

##### transition\_matrix

[DataFrame] (column) channel\_from: channel from (column) channel\_to : channel to (column) transition\_probability : transition probability from channel\_from to channel\_to

##### removal\_effects:

(column) channel\_name : channel names (column) removal\_effects\_conversion : removal effects for each channel calculated using total conversions (column) removal\_effects\_conversion\_value : removal effects for each channel calculated using revenues

#### Return type

list of DataFrames

#### Examples

Load Data

```
>>> import pandas as pd
>>> from ChannelAttribution import *
>>> Data = pd.read_csv('https://channelattribution.io/csv/Data.csv', sep=";")
```

Estimate an automatic Markov model

```
>>> auto_markov_model(Data, "path", "total_conversions", "total_null")
```

ChannelAttribution.**choose\_order**(Data, var\_path, var\_conv, var\_null, max\_order=10, sep='>', ncore=1, roc\_npt=100, plot=True, flg\_pro=True)

Find the minimum Markov Model order that gives a good representation of customers' behaviour for data considered. It requires paths that do not lead to conversion as input. Minimum order is found maximizing a penalized area under ROC curve.



**Parameters**

- **Data** (*DataFrame*) – customer journeys.
- **var\_path** (*string*) – column of Data containing paths.
- **var\_conv** (*string*) – column of Data containing total conversions for each path.
- **var\_null** (*string*) – column of Data containing total paths that do not lead to conversion.
- **max\_order** (*int*, *default 10*) – maximum Markov Model order to be considered.
- **sep** (*string*, *default ">"*) – separator between the channels.
- **ncore** (*int*, *default 1*) – number of threads to be used in computation.
- **roc\_npt** (*int*, *default 100*) – number of points to be used for the approximation of roc curve.
- **plot** (*bool*, *default True*) – if True, a plot with penalized auc with respect to order will be displayed.
- **flg\_pro** (*bool*, *default True*) – if True, ChannelAttribution Pro banner is printed.

**Returns****roc**

[list DataFrame one for each order considered] (column) tpr: true positive rate (column) fpr: false positive rate

**auc**

[DataFrame with the following columns] (column) order: markov model order (column) auc: area under the curve (column) pauc: penalized auc

**suggested order**

[int] estimated best order

**Return type**

list

**Examples**

Estimate best markov model order for your data

Load Data

```
>>> import pandas as pd
>>> from ChannelAttribution import *
>>> Data = pd.read_csv('https://channelattribution.io/csv/Data.csv', sep=";")
```

```
>>> choose_order(Data, var_path="path", var_conv="total_conversions", var_null=
↳ "total_null")
```

`ChannelAttribution.heuristic_models(Data, var_path, var_conv, var_value=None, sep='>', flg_pro=True)`

Estimate three heuristic models (first-touch, last-touch and linear) from customer journey data.

**Parameters**

- **Data** (*DataFrame*) – customer journeys.
- **var\_path** (*string*) – column of Data containing paths.
- **var\_conv** (*string*) – column of Data containing total conversions for each path.

- **var\_value** (*string, optional, default None*) – column of Data containing revenue for each path.
- **sep** (*string, default ">"*) – separator between the channels.
- **flg\_pro** (*bool, default True*) – if True, ChannelAttribution Pro banner is printed.

**Returns**

(column) **channel\_name** : channel names (column) **first\_touch\_conversions** : conversions attributed to each channel using first touch attribution. (column) **first\_touch\_value** : revenues attributed to each channel using first touch attribution. (column) **last\_touch\_conversions** : conversions attributed to each channel using last touch attribution. (column) **last\_touch\_value** : revenues attributed to each channel using last touch attribution. (column) **linear\_touch\_conversions** : conversions attributed to each channel using linear attribution. (column) **linear\_touch\_value** : revenues attributed to each channel using linear attribution.

**Return type**

DataFrame

**Examples**

Load Data

```
>>> import pandas as pd
>>> from ChannelAttribution import *
>>> Data = pd.read_csv('https://channelattribution.io/csv/Data.csv', sep=";")
```

Estimate heuristic models on total conversions

```
>>> heuristic_models(Data, "path", "total_conversions")
```

Estimate heuristic models on total conversions and total revenues

```
>>> heuristic_models(Data, "path", "total_conversions", \
>>> var_value="total_conversion_value")
```

**ChannelAttribution.install\_pro()**

Interactive installer for ChannelAttribution Pro (Python wheel).

This helper: - Prompts you for a token; if you type an email instead, it requests a token for that address and exits (so you can rerun with the token).

- Detects your OS/arch/Python, asks the build service for the correct wheel via HTTPS POST, resolves the final wheel URL, and installs it with `python -m pip`.
- Sends exactly one final outcome notification to `build_check_email.php` (SUCCESS/ERROR) with a short, client-trimmed info blob.

**Notes**

User interaction: - On start, it asks: “Enter your ChannelAttributionPro token. If you don’t have one, enter your work/university email to request it:”

- If you enter a valid email, the function triggers a token email and returns immediately.
- If you enter a non-empty token, the installation flow proceeds.
- If you enter nothing, the function prints a message and returns.

Environment detection: - OS name: manylinux (Linux), macOS (macOS), Windows (Windows) - OS version: Linux “2014” (ManyLinux2014), macOS “13” (amd64) or “15” (arm64), Windows “11” - Architecture: amd64 (x86\_64) or arm64 (aarch64) - Python: running interpreter’s major.minor (e.g., 3.11)

Network behavior: - Builder request: HTTPS POST to <https://app.channelattribution.io/genpkg/genpkg.php>

with {os, os\_vers, arch, lang=python, lang\_vers, replace=0, uctr=0, token}.

- Response handling: - 200 or 409 with JSON containing "pkg": direct wheel URL or a directory URL. - If a directory is returned, the latest file is selected (prefers \*.whl). - 401 or JSON error mentioning “invalid token”: prints “Token non valid or expired...” and returns.
- Installation: runs `python -m pip install --prefer-binary` on the resolved wheel URL.
- Final notification (always exactly once): POSTs to [https://app.channelattribution.io/genpkg/build\\_check\\_email.php](https://app.channelattribution.io/genpkg/build_check_email.php) with fields {token, action="SUCCESS"|"ERROR", info="<=8KiB compact JSON-like"}.

Packages and proxies: - Auto-installs `requests` (and `distro` on Linux) into the current interpreter.

If not in a virtualenv, sets `PIP_BREAK_SYSTEM_PACKAGES=1` to respect Debian/Ubuntu PEP 668.

- Honors standard system/proxy environment variables.

Output and errors: - Prints progress and a short system report if something goes wrong (OS, distro, Python, compiler).

Support message includes [info@channelattribution.io](mailto:info@channelattribution.io).

- On success it prints:

```
*** Package installed. Restart the session and try to import it with: import_
↳ ChannelAttributionPro
```

- Returns `None`. Most error conditions are handled by printing and returning.
- May raise `ValueError` only if an email-like input is given but fails basic syntax validation.

Security notes: - Sends your token to the builder and in the final notification. - The `info` field is trimmed client-side (~8 KiB) and HTML-escaped server-side before emailing. - No personal data is sent beyond the token and generic environment traits.

## Examples

```
>>> from ChannelAttribution import install_pro
>>> install_pro() # follow the prompt: paste token, or enter email to request one
```

```
ChannelAttribution.markov_model(Data, var_path, var_conv, var_value=None, var_null=None, order=1,
                                nsim_start=100000.0, max_step=None, out_more=False, sep='>',
                                ncore=1, nfold=10, seed=0, conv_par=0.05, rate_step_sim=1.5,
                                verbose=True, flg_pro=True)
```

Estimate a k-order Markov model from customer journey data. Differently from `markov_model`, this function iterates estimation until a desired convergence is reached and enables multiprocessing.

### Parameters

- **Data** (`DataFrame`) – customer journeys.

- **var\_path** (*string*) – column of Data containing paths.
- **var\_conv** (*string*) – column of Data containing total conversions for each path.
- **var\_value** (*string, optional, default None*) – column of Data containing revenue for each path.
- **var\_null** (*string*) – column of Data containing total paths that do not lead to conversion.
- **order** (*int, default 1*) – Markov model order.
- **nsim\_start** (*int, default 1e5*) – minimum number of simulations to be used in computation.
- **max\_step** (*int, default None*) – maximum number of steps for a single simulated path. if NULL, it is the maximum number of steps found into Data.
- **out\_more** (*bool, default False*) – if True, transition probabilities between channels and removal effects will be returned.
- **sep** (*string, default ">"*) – separator between the channels.
- **ncore** (*int, default 1*) – number of threads to be used in computation.
- **nfold** (*int, default 10*) – how many repetitions to be used to verify if convergence has been reached at each iteration.
- **seed** (*int, default 0*) – random seed. Giving this parameter the same value over different runs guarantees that results will not vary.
- **conv\_par** (*double, default 0.05*) – convergence parameter for the algorithm. The estimation process ends when the percentage of variation of the results over different repetitions is less than convergence parameter.
- **rate\_step\_sim** (*double, default 0*) – number of simulations used at each iteration is equal to the number of simulations used at previous iteration multiplied by rate\_step\_sim.
- **verbose** (*bool, default True*) – if True, additional information about process convergence will be shown.
- **flg\_pro** (*bool, default True*) – if True, ChannelAttribution Pro banner is printed.

## Returns

### result: Dataframe

(column) channel\_name : channel names (column) total\_conversions : conversions attributed to each channel (column) total\_conversion\_value : revenues attributed to each channel

### transition\_matrix

[DataFrame] (column) channel\_from: channel from (column) channel\_to : channel to (column) transition\_probability : transition probability from channel\_from to channel\_to

### removal\_effects:

(column) channel\_name : channel names (column) removal\_effects\_conversion : removal effects for each channel calculated using total conversions (column) removal\_effects\_conversion\_value : removal effects for each channel calculated using revenues

## Return type

list of DataFrames

## Examples

Load Data

```
>>> import pandas as pd
>>> from ChannelAttribution import *
>>> Data = pd.read_csv('https://channelattribution.io/csv/Data.csv', sep=";")
```

Estimate a Markov model using total conversions

```
>>> markov_model(Data, "path", "total_conversions")
```

Estimate a Markov model using total conversions and revenues

```
>>> markov_model(Data, "path", "total_conversions", var_value="total_conversion_
↪value")
```

Estimate a Markov model using total conversions, revenues and paths that do not lead to conversions

```
>>> markov_model(Data, "path", "total_conversions", var_value="total_conversion_
↪value", var_null="total_null")
```

Estimate a Markov model returning transition matrix and removal effects

```
>>> markov_model(Data, "path", "total_conversions", var_value="total_conversion_
↪value", var_null="total_null", out_more=True)
```

Estimate a Markov model using 4 threads

```
>>> markov_model(Data, "path", "total_conversions", var_value="total_conversion_
↪value", ncore=4)
```

`ChannelAttribution.transition_matrix(Data, var_path, var_conv, var_null, order=1, sep='>',  
flg_equal=True, flg_pro=True)`

Estimate a k-order transition matrix from customer journey data.

### Parameters

- **Data** (*DataFrame*) – customer journeys.
- **var\_path** (*string*) – column of Data containing paths.
- **var\_conv** (*string*) – column of Data containing total conversions for each path.
- **var\_null** (*string*) – column of Data containing total paths that do not lead to conversion.
- **order** (*int*, *default 1*) – Markov model order.
- **sep** (*string*, *default ">"*) – separator between the channels.
- **flg\_equal** (*bool*, *default True*) – if True, transitions from a channel to itself will be considered.
- **flg\_pro** (*bool*, *default True*) – if True, ChannelAttribution Pro banner is printed.

### Returns

**channels:** *Dataframe*

(column) `id_channel` : channel ids (column) `channel_name` : channel names

**transition\_matrix**

[DataFrame] (column) channel\_from: id channel from (column) channel\_to : id channel to  
(column) transition\_probability : transition probability from channel\_from to channel\_to

**Return type**

list of DataFrames

**Examples**

Load Data

```
>>> import pandas as pd
>>> from ChannelAttribution import *
>>> Data = pd.read_csv('https://channelattribution.io/csv/Data.csv', sep=";")
```

Estimate a second-order transition matrix using total conversions and paths that do not lead to conversion

```
>>> transition_matrix(Data, "path", "total_conversions", var_null="total_null",
↳ order=2)
```

## PYTHON MODULE INDEX

### C

`ChannelAttribution`, [3](#)





## INDEX

### A

`auto_markov_model()` (*in module ChannelAttribution*),  
3

### C

`ChannelAttribution`  
module, 3

`choose_order()` (*in module ChannelAttribution*), 4

### H

`heuristic_models()` (*in module ChannelAttribution*),  
5

### I

`install_pro()` (*in module ChannelAttribution*), 6

### M

`markov_model()` (*in module ChannelAttribution*), 7  
module  
    `ChannelAttribution`, 3

### T

`transition_matrix()` (*in module ChannelAttribution*),  
9