Multi-Channel Attribution Modeling With AuriQ Essentia



Essentia Marketing Attribution Whitepaper

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Marketing Attribution with AuriQ ESSENTIA

A Fast, Flexible and (user) Friendly way to get the complete and accurate description of the customer journey from impressions to conversion

With ESSENTIA, marketing attribution modeling has never been so easy. ESSENTIA allows you to link impressions, clicks, organic and paid searches, and conversion events to get the most complete and accurate description of the customer journey experience. With ESSENTIA, merging multiple sizable logs and computing attribution scores becomes an easy task.



ESSENTIA : The User Experience

From raw MediaMind logs to attribution scores

Computing attribution scores with ESSENTIA is a straightforward and easy process. Here we provide a brief ESSENTIA user cookbook to get you from raw MediaMind logs to a csv table of attribution scores.

1. Customize ESSENTIA to fit your needs

ESSENTIA provides a library of attribution models with flexible parameters. The first step in computing attribution scores is to customize and tailor the ESSENTIA config_file to fit your needs. This step involves :

- a. specifying your conversion, standard and AQ (optional) log files;
- b. specifying the conversion tag ID's and campaign ID's of interest;
- c. specifying the **type of attribution model** you want to compute along with its associated parameters;
- d. the name of file containing the final scores.

We provide **an example of a configuration file in Appendix A**

2. Load the log data in ESSENTIA

Load the MM log data into ESSENTIA using this command :

% import_and_match config_file

ESSENTIA will grab the **compressed** files and filter the conversion, standard and AQ (optional) log data according to the parameters specified in the config_file. Different log entries are joined based on user ID's. **Essentia will generate a complete an accurate timeline of the customer journey for each customer who converted.** If AQ logs are specified, ESSENTIA will join the AQ log entries based on a combination of event timestamp and user IP address. The data are now cleaned, filtered and loaded into memory. The next step is to compute attribution scores.



3. Run ESSENTIA powerful attribution modeling tools

Once the data have been successfully imported in ESSENTIA, you can run attribution models by way of a simple command line :

% attribution_model config_file

ESSENTIA provides a variety of attribution models including:

- a. **linear** : all impressions, clicks and searches in the customer journey are credited in the attribution. The user can specify the relative weight of each event in the config_file as well as a time window before conversion;
- b. **first click** : the first click in the customer journey (and within a given time window before conversion) gets all the credit for the conversion;
- c. last click : in this case, the last click gets all the credit;
- d. **first interaction** : the first interaction gets all the credit regardless of the type. This first interaction could be an impression, a click, a search, etc. ;
- e. **last interaction** : in this case, the last interaction gets all the credit;
- f. **time decay** : each step in the customer journey is weighted can specify the half-life parameter of the function as well as the relative weight of each type of event/interaction.

Want to compare the output of different models? That's EASY and FAST with ESSENTIA. Simply re-run the above attribution_model command with a different config_file. Since the customer journey data are still loaded in memory, ESSENTIA allows you to quickly and efficiently run multiple attribution models in a serial fashion. Your favorite model is not listed? Let us know and we will customize ESSENTIA attribution models to fit your needs.

4. Terminate

Once you are satisfied with your attribution scores, simply enter :

<mark>% udbd stop</mark>

The memory block containing the customer journey data will be freed



Attribution Scores and Beyond

Go beyond simple attribution scores... the easy way

ESSENTIA produces a series of files in csv and txt formats (ideal for your favorite analysis tool such as EXCEL, Tableau, Qlikview, etc) containing attribution scores as well as critical metadata about the customers journeys from impression all the way through conversion.

Here are just a few examples of analytics performed with ESSENTIA's outputs



Compare your favorite models

Not sure if a linear attribution is the best? How about first click or time decay? With ESSENTIA, different attribution models results can be compared easily.



Time Evolution

How do attribution scores change over time? With ESSENTIA, you can compare attribution scores over daily, weekly, bi-weekly, monthly or yearly baselines.



Conversion Metadata

How does the mean path length to conversion change over time? Did the number of conversions increase this past week?. With ESSENTIA attribution tools, these questions and many more can be easily answered.

Beyond Attribution: A Network Topology Approach to the Customer Journey

ESSENTIA allows you to export conversion paths to your favorite network graph visualization tool (e.g. Gephi) to explore the network topology of customer journeys and the links in your marketing strategy

Sample Results: A MediaMind Client Get actionable results faster

Linear attribution scores over a one-day period. Top 14 results shown Comparison between multiple attribution models





Attribution Scores - Comparison between Different Models

A Network Topology Approach to the Customer Journey

Get a graphical representation of the customers' journeys from impressions through conversion. Explore the links between touch points with your favorite network graph software (e.g. Gephi)



Conversion Metadata: time evolution of the number of conversions and mean path length over a week period



Conversion Metadata: following your customers over 10 days of impressions, clicks, and organic search data



For this Campaign, all the action... in the last 3 hours before conversion! (each bar represents 1 hour)



Appendix A - Example of a config file

Specify the logs here. Note that you can use * or ? in the name of the file if # more than one file is included. Note that AQ Logs are optional. ConversionLog ~ data/MM CLD Conversion Advertiser 65303 Daily 130601.zip StandardLog ~/data/MM CLD Standard Advertiser 65303 Daily 130601.zip AQLog ~/data/skp130601.csv.zip # Specify the converion Tag IDs of interest here : ConversionTagID 169841 169843 # Specify the campaign IDs of interest here : CampaignID 309302 # Specify the EntityID's you want to select OR exclude. # By default, all EntityIDs are considered # if you want to exclude IDs<mark>,</mark> put an "E" in front of the IDs. # examples : # you want to focus solely on Entity IDs 13689119, 14035878, and 14036626 # then put : # EntityID 13689119 14035878 14036626 # you want to exclude these very same IDs from the attribution scores : # EntityID E13689119 E14035878 E14036626 # Note that you can't exclude and include IDs in the same EntityID line EntityID E14028158 E14035903 E13689119 E14035878 E14036599 E14036626 E14036711 E14036770 E14036690 E14039370 E14036663 E14039239 # Attribution Modeling Parameters : # Select *ONE* model. Default is linear with w imp=1.0, w click=1.0, w sem=0, # w seo=0. # Attribution models currently available are : .:. linear : all events in the customer journey are considered # (including impression) The relative weight of each event is determined by the w * parameters below. # The first click gets all the credit for the .:. first click : # conversion. # .:. last click : The last click gets all the credit for the conversion. # # .:. first interaction : The first interaction/event in the customer journey # gets all the credit for the conversion, regardless of the type of the interaction/event. # # # In this case, the last interaction gets all the .:. last interaction : # credit. # # .:. time decay : The weight of each event/interaction follows an # exponentially declining function characterized by

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#
                              the parameter half-life (see below). Note that
#
                              the user can specify the relative weight of each type
#
                              of events (w_*). These parameters will MODULATE the
#
                              time decay weights. If you want a set of weights
                              purely determined by a time-decay exponential, set
#
#
                              all w * parameters to 1.0.
#
Attribution model linear
# output file containing the scores. Default is am.out
output file am.out
# Time window before the first conversion. Weights are assign to events falling
# within this time window. All events prior to the time window are discarded (i.e.
# all events at time < (conversion time - time window))</pre>
# Default of time window is 86400 sec (= 1day)
time window 86400
# Relative weights of different types of interactions/events :
# Impressions. default value for w imp = 1 (except in the case of * click models,
w imp=0)
w imp 1.0
# Clicks. default value for w click = 1 (if w click is not specified)
w click 1.0
# Organic searches . default value of w seo = 0 (if w seo is not specified)
w seo 1.0
# Search-engine marketing (SEM) - Paid searches. default value of w sem = 0
# (if w sem is not specified).
w sem 1.0
# if using timedecay model, specify the half-life :
# default value is half life = 86400 (=1day)
half life 86400
```



Want to learn more about AuriQ Essentia's big log processing capabilities? Do you have a project in mind already?

Please contact us for a free consultation. We'll assess your data analysis needs and help you determine if AuriQ Essentia is right for you.

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