

Customer live time value

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Marketing One aim of marketing is to establish a "good" relationship to the company's customers. In terms of brutal capitalism this is equivalent to maximizing a customer's contribution to profit of the company. From a marketing perspective the estimate of this quantity must also include the contribution of a customer's future contribution to profit. This is because a marketing guy needs to decide whether or not a customer is worth participating in a certain marketing campaign (e.g., SEA, sending a voucher, catalogue or email, etc.). Thus, the challenge is to estimate this quantity over the whole "life time" of a customer. This quantity is called customer live time value (CLV). So we can say, marketing is concerned about maximising the CLV and customer equity (i.e., the sum of the life time values of the company's customers).

Operational Implications It is crucial to recall that the CLV is an estimate given the customer's transaction history. So, historically his shopping behavior is influenced by historical marketing actions. Furthermore, the CLV has empirically always an exponential decay. This implies that customers will die (either physically or economically) sooner or later. So one could use the CLV to monitor marketing actions and pose the question: did this marketing action prolong the customer's life time?

Categorization Modelling the customer live time value (CLV) has quite a long history in direct marketing [9, 2, 6]. There exist Bayesian approaches and extensions [7, 8] as well as more machine learning based approaches [1]. In the conventional setting the different models that exist can be categorized by contractual vs non-contractual and continuous purchases vs discrete purchases settings [5]. A contractual setting is a business model with contract (e.g., newspaper subscription). A continuous setting is a business model where buy events may happen at any point in time (e.g., amazon).

	Non-contractual	Contractual
Continuous	<ul style="list-style-type: none">• Use case: Amazon• Pareto/NBD model [9]• $\Gamma - \Gamma$ extension [3, 4]• BG/NBD [2]	<ul style="list-style-type: none">• Use case: Credit Card
Discrete	<ul style="list-style-type: none">• Use case: Event Attendance	<ul style="list-style-type: none">• Use case: Netflix

Note that conventional CLV models [9] focus on purchase counts and life time. The $\Gamma - \Gamma$ extension [4] is a method to also incorporate monetary value.

1 The beta/geometric beta-Bernoulli model

[2]

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