Loan Pricing Modeling Notes

Last update

August 20, 2020

Risk Based Pricing

1.1 Preliminary definitions

We will start by characterizing the payouts for a constant installment loan under the assumption that there is no risk. In this world, there is only the contractual cash-flow and no risk of any kind (e.g. prepayment risk, default risk,etc). In section (1.3) we will introduce those risk factors from first principles. ¹

The connection between the risk free world and the real world will be established through the theorem of telescopic amortizations in section (1.3). To the best of my knowledge, this theorem has not been established anywhere else but it turns out to be extremely useful for two reasons: First, using this theorem we can easily stablish a closed form mathematical representation of the Economic Profit of a loan from first priciples even allowing for different modeling choices for the sequence of computations.

Second, the use of the theorem allow for such a simple representation that its programatic implementation does not need any for-loops of any kind, freeing the space for using the vectorization approach that is widely used in data intensive applications in statistics and deep learning.

Element	Notation
Client Interest Rate	r
Cost of Funds Rate, Fund Transfer Pricing FTP, TT	r_c
Discount Rate	r_d
Contractual Maturity	T

¹The reader familiarized with modern mathematical asset pricing will recognize the similarity of this approach with the typical Q and P measure approach, - i.e. studying first the asset price under a risk neutral world (measure Q) and then connecting with the real world which is non-risk-neutral (P measure)

1.2 Payment schedule in a risk free world

1.2.1 The Constant Installment Function

We can define a $c_f(.)$ function to compute constant installments by defining the following equality involving c (the constant installments), $\bar{B}(1)$ (the lended principal), r (the loan interest rate) and T the contractual maturity.

$$\bar{B}(1) = \frac{c}{(1+r)} + \frac{c}{(1+r)^2} + \frac{c}{(1+r)^3} + \dots + \frac{c}{(1+r)^T}$$
(1.1)

Setting $\delta = \frac{1}{1+r}$

$$\bar{B}(1) = c\delta \left[1 + \delta + \delta^2 + \delta^3 + \delta^4 + \dots + \delta^{T-1}\right]$$

$$= c\delta \left[\frac{1}{1 - \delta} - \delta^T \frac{1}{1 - \delta}\right] = c\delta \left(\frac{1 - \delta^T}{1 - \delta}\right)$$
(1.2)

Solving for c we stablish the definition for $c_f(r,T)$

$$c = \bar{B}(1) \left(\frac{1-\delta}{\delta}\right) \frac{1}{1-\delta^T} = B(1) \left[r \frac{(1+r)^T}{(1+r)^T - 1}\right]$$
(1.3)

$$c_f(r,T) := \frac{r(1+r)^T}{(1+r)^T - 1} \implies c = \bar{B}(1)c_f(r,T)$$
(1.4)

1.2.2 The Balance Function:

The Current Balance function $\bar{B}(t)$ is the remaining balance left at time t-1 for a loan with principal $B = \bar{B}(1)$ and in absence of any prepayment or default risk ². For a constant installments loan, the remaining balance at time t for a loan with principal (Balance at t=0) B is given by

$$\bar{B}(t) = B \frac{(1+r)^T - (1+r)^{t-1}}{(1+r)^T - 1}$$
(1.5)

To establish equation (1.16) we start from first principles noticing that the remaining balance is the capitalized previous balance minus the installment

$$\bar{B}(t) = \bar{B}(t-1)(1+r) - c \tag{1.6}$$

Equation (1.6) is a difference equation of first order that can be solved recursively with initial value given by the principal B as follows:

²The acute reader will notice that the definition of $\bar{B}(t)$, in terms of the remaining balance left at time t-1, was given so that we can state such a simple equation for the earned contractual interest $\bar{I}(t) = r\bar{B}(t)$ and eliminate dependence on past indexes t.

$$\bar{B}(1) = B
\bar{B}(2) = \bar{B}(1)(1+r) - c = \bar{B}(1+r) - c
\bar{B}(3) = \bar{B}(2)(1+r) - c = \bar{B}(1+r)^2 - c(1+r) - c
\bar{B}(4) = \bar{B}(3)(1+r) - c = \bar{B}(1+r)^3 - c(1+r)^2 - c(1+r) - c
\bar{B}(5) = \bar{B}(4)(1+r) - c = \bar{B}(1+r)^4 - c(1+r)^3 - c(1+r)^2 - c(1+r) - c$$

$$\bar{B}(t) = \bar{B}(1+r)^{t-1} - c\sum_{s=0}^{t-2} (1+r)^s$$
(1.7)

Using the fact that: If $S_n = 1 + x + x^2 + x^3 + ... + x^n \implies S_n = \frac{1 - x^{n+1}}{1 - x}$

$$\bar{B}(t) = \bar{B}(1+r)^{t-1} - c\left[\frac{(1+r)^{t-1} - 1}{r}\right]$$
(1.8)

Replacing (1.4) in (1.8)

$$\bar{B}(t) = \bar{B}(1) \left[\frac{(1+r)^T - (1+r)^{t-1}}{(1+r)^T - 1} \right]$$
(1.9)

$$B_f(t,r,T) = \left[\frac{(1+r)^T - (1+r)^{t-1}}{(1+r)^T - 1} \right] \implies \bar{B}(t) = \bar{B}(1)B_f(t,r,T)$$
 (1.10)

1.2.3 Amortization factor:

Finally, we define the amortization factor noticing the constant installment should be equal to the interest payment plus the amortization.

$$A_f(t, r, T) = c_f(r, T) - rB_f(t, r, T)$$
(1.11)

$$A_f(t, r, T) = \frac{r(1+r)^{t-1}}{(1+r)^T - 1}$$
(1.12)

Corollary 1. Using (1.6) and (1.11) we can state that:

$$1 - \sum_{s=1}^{t-1} A_f(s, r, T) = B_f(t)$$

1.3 Introducing Risk Events in Payment schedule

Last section we defined financial math operators to define a loan payment schedule in absence of any risk events. In this section we introduce risk events including defaults and prepayments. Lets start with the Theorem of Telescopic Amortizations which will be key to connect the risk free schedule with the risky schedule.

Theorem 1.1 (Telescopic Amortizations). If A is the function defined in (1.12) then:

$$\prod_{s=1}^{t-1} (1 - A_f(1, r, T - s + 1)) = 1 - \sum_{s=1}^{t-1} A_f(s, r, T) = B_f(t, r, T)$$

Proof. Lets define

$$E_1 = \prod_{s=1}^{t-1} (1 - A_f(1, r, T - s + 1)), E_2 = 1 - \sum_{s=1}^{t-1} A_f(s, r, T)$$

and

$$\delta = 1/(1+r)$$

Working on E_1 and setting $\xi = 1 + r$:

$$1 - A_f(1, r, T - s + 1) = \frac{(1+r)^{T-s+1} - 1 - r}{(1+r)^{T-s+1} - 1} = \frac{\xi^{T-s+1} - \xi}{\xi^{T-s+1} - 1}$$

$$\implies E_{1} = \frac{(\xi^{T} - \xi)}{(\xi^{T} - 1)} \times \frac{(\xi^{T-1} - \xi)}{(\xi^{T-1} - 1)} \times \frac{(\xi^{T-2} - \xi)}{(\xi^{T-2} - 1)} \times \dots \times \frac{(\xi^{T-t+1} - \xi)}{(\xi^{T-t+1} - 1)}$$

$$= \xi^{t} \frac{(\xi^{T-1} - 1)}{(\xi^{T} - 1)} \times \frac{(\xi^{T-2} - 1)}{(\xi^{T-1} - 1)} \times \frac{(\xi^{T-3} - 1)}{(\xi^{T-2} - 1)} \times \dots \times \frac{(\xi^{T-t+1} - 1)}{(\xi^{T-t+1} - 1)}$$

$$= \frac{\xi^{T} - \xi^{t}}{\xi^{T} - 1}$$

$$(1.13)$$

Working on E_2 :

$$A_f(s, r, T) = \frac{r(1+r)^{s-1}}{(1+r)^T - 1} = (\xi - 1)\frac{\xi^{s-1}}{\xi^T - 1}$$

$$\implies E_2 = 1 - \frac{(\xi - 1)(1 + \xi + \xi^2 + \xi^3 + \dots + \xi^{t-1})}{\xi^T - 1}$$

$$= \frac{\xi^T - \xi^t}{\xi^T - 1}$$
(1.14)

 $\therefore E1 = E2$

Finally, the second equality follows directly from Corollary 1.

1.3.1 Default and Prepayment probabilities and the survival function:

We define de conditional probabilities $p_p(t)$ and $p_d(t)$ $\forall t \in \{1, 2, ... T\}$ as follows:

 $p_p(t)$: Probability that the loan will prepay at time t given that it has survived to that point $p_d(t)$: Probability that the loan will default at time t given that it has survived to that point

Given these definitions we can establish S(t), the probability that a loan survives until period t using the pigeon hole priciple.

$$S(t) = (1 - p_d(1) - p_p(1)) \times (1 - p_d(2) - p_p(2)) \times \dots \times (1 - p_d(t) - p_p(t))$$

= $\prod_{s=1}^{t} (1 - p_d(s) - p_p(s))$ (1.15)

1.4 Alternative setups for Incremental Profit computation

1.4.1 Standard model

In this model prepayments/full prepayment, default probability are expressed as conditional probabilities. This probabilities are conditioned on the running active balance i.e. the balance that has not been prepaid or defaulted upon.

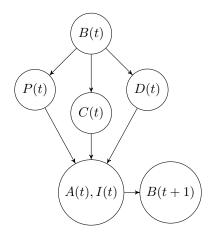


Figure 1.1: Computation Graph

Variable	Notation	Calculation
Balance in presence of risk	B(t)	B(t)
Default	D(t)	$p_d(t)B(t)$
Full Prepayment	C(t)	$p_c(t)B(t)$
Prepayment	P(t)	$p_p(t)B(t)$
Amortization	A(t)	$B(t) - D(t) - C(t) - P(t)A_f(1, r, T - t + 1)$
Interest	I(t)	(B(t) - D(t) - C(t) - P(t))r
Principal	B(1)	B
Risky balance next period	B(t+1)	B(t) - D(t) - C(t) - P(t) - A(t)

Table 1.1: Computation for Model 1: Standard Model

Notice that we defined B(t) as the Loan Balance subject to risk (Conductual affected Loan Balance), as opposed to $\bar{B}(t)$, which is the Risk Free Loan Balance (Contractual Balance). Given this definitions we can compute the recursive form for the balance function B(t)

$$B(t+1) = B(t)[1 - p_d(t) - p_c(t) - p_p(t) - (1 - p_d(t) - p_c(t) - p_p(t))A_f(1, r, T - t + 1)]$$

= $B(t)(1 - p_d(t) - p_c(t) - p_p(t))(1 - A_f(1, r, T - t + 1))$ (1.16)

Notice that (1.16) is a first order equation in difference which can be easily solved as.

$$B(t) = \prod_{s=1}^{t-1} (1 - A_f(1, r, T - s + 1))(1 - p_d(s) - p_c(s) - p_p(s))B$$
 (1.17)

Using theorem (1.3) we can state the conductual balance B(t) as a function of the contractual balance.

$$B(t) = \prod_{s=1}^{t-1} (1 - p_d(s) - p_c(s) - p_p(s)) \bar{B}(t)$$

$$= S(t-1)\bar{B}(t)$$

$$B(t) = S(t-1)\bar{B}(t)$$
(1.18)

Equation 1.18 states a very simple relation between the theoretical/contractual balance $\bar{B}(t)$ and the behavioral balance B(t). We have derived this equation from a first principles approach through the Theorem of Telescopic Amortization (1.3). This equation represent a very powerful shortcut not only for using intuition, since the behavioral balance can be thought of as the contractual balance adjusted by the survival probability, but also for implementing the model programatically using vectorization instead of recursive loops over the different points in the payment schedule, the last alternative can be very hard to maintain and compute, not to mention its proneness to error.

Corollary: Interest in model 1. Given the interest rate definition in Table (1.1) and equation (1.18) we can state that:

$$I(t) = S(t)\bar{B}(t)r \tag{1.19}$$

1.4.2 Prepayment dependent on initial balance

This model is a variation of the previous one in which the prepayment probability is a proportion of the initial balance/principal B(1). In this setup, the prepayment amount will be define in function of the marginal probability of default $p_p^m(t)$ and not the conditional probability $p_p(t)$

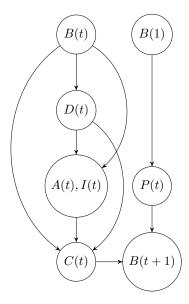


Figure 1.2: Computation Graphs: Model 2

Variable	Notation	Calculation
Balance in presence of risk	B(t)	B(t)
Default	D(t)	$p_d(t)B(t)$
Amortization	A(t)	$(B(t) - D(t))A_f(1, r, T - t + 1)$
Interest	I(t)	(B(t) - D(t))r
Full Prepayment	C(t)	$p_c(t)[B(t) - D(t) - A(t)]$
Principal	B(1)	B
Prepayment	P(t)	$p_p^m(t)B$
Risky balance next period	B(t+1)	B(t) - D(t) - C(t) - P(t) - A(t)

Table 1.2: Computation for Model 2: Prepayment Dependent on Initial Balance

Given this definitions we can compute the recursive form for the balance function B(t)

$$B(t+1) = B(t)[1 - p_d(t) - p_c(t)(1 - p_d(t) - (1 - p_d(t))A_f(1, r, T - t + 1)) - (1 - p_d(t))A_f(1, r, T - t + 1)] - p_p^m(t)B(1)$$

$$= B(t)[1 - p_d(t) - p_c(t) + p_d(t)p_c(t) + p_c(t)(1 - p_d(t)A_f(1, r, T - t + 1) - (1 - p_d(t))A_f(1, r, T - t + 1)] - p_p^m(t)B(1)$$

$$= B(t)[(1 - p_d(t))(1 - p_c(t)) - (1 - p_d(t))(1 - p_c(t))A_f(1, r, T - t + 1)] - p_p^m(t)B(1)$$

$$= B(t)(1 - p_d(t))(1 - p_c(t))(1 - A_f(1, r, T - t + 1)) - p_p^m(t)B(1)$$

$$(1.20)$$

Notice that (1.20) is a first order equation in difference of type $x(t+1) = \alpha(t)x(t) + \beta(t)$ where $\alpha(t) := (1-p_d(t))(1-p_c(t))(1-A_f(1,r,T-t+1))$, $\beta(t) := -p_p^m(t)\bar{B}(1)$, $x(t) := \bar{B}(t)$ and the initial condition given by the lended principal $x(1) = \bar{B}(1) = B$. This equation can be easily solved as:

$$x(t) = x_1 \prod_{s=1}^{t-1} \alpha(s) + \sum_{k=1}^{t-2} \left[\beta(k) \prod_{s=k+1}^{t-1} \alpha(s) \right] + \beta(t-1)$$
 (1.21)

Plugging back the definitions of $\alpha(t)$, $\beta(t)$ and x(t) we get:

$$B(t) = \prod_{s=1}^{t-1} (1 - p_d(s))(1 - p_p(s))(1 - A_f(1, r, T - s + 1))B$$

$$-\sum_{k=1}^{t-2} \left[p_p^m(k)B \prod_{s=k+1}^{t-1} (1 - p_d(s))(1 - p_c(s))(1 - A_f(1, r, T - s + 1)) \right] - p_p^m(t - 1)B$$
(1.22)

Using the theorem of Telescopic Amortizations (1.3) and defining $\tilde{S}(t) := \prod_{s=0}^{t} (1 - p_d(s))(1 - p_c(s))$ we can state the behavioral balance B(t) as a function of the contractual balance $\bar{B}(t)$.

$$\begin{split} B(t) &= \tilde{S}(t-1)\bar{B}(t) - \sum_{k=1}^{t-2} \left[p_p^m(k) B \frac{\tilde{S}(t-2)}{\tilde{S}(k)} \frac{\bar{B}(t-1)}{\bar{B}(k+1)} \right] - p_p^m(t-1) B \\ &= \tilde{S}(t-1)\bar{B}(t) - \sum_{k=1}^{t-1} \left[p_p^m(k) B \frac{\tilde{S}(t-1)}{\tilde{S}(k)} \frac{\bar{B}(t)}{\bar{B}(k+1)} \right] \\ &= \tilde{S}(t-1) \left(1 - \sum_{k=1}^{t-1} \frac{p_p^m(k) B}{\tilde{S}(k) \bar{B}(k+1)} \right) \bar{B}(t) \end{split}$$

We define conveniently S(t) as:

$$S(t-1) := \tilde{S}(t-1) \left(1 - \sum_{k=1}^{t-1} \frac{p_p^m(k)B}{\tilde{S}(k)\bar{B}(k+1)} \right)^+$$
 (1.23)

So that we can state that:

$$B(t) = S(t-1)\bar{B}(t)$$
(1.24)

Equation (1.23) is analogous to equation (1.18) with an additional term that represents a weighted sum of the last prepayment amounts where the last prepayment has a weight of one.

1.5 Terms included in the incremental profit (CLV)

1.5.1 Interest on loans:

$$LI(t) = S(t)\bar{B}(t)r \tag{1.25}$$

Equation (1.25) will be proved in section (1.4.1). For now lets just use our intuition and state that each dollar has an unconditional probability to survive up to time t of S(t)

1.5.2 Cost of Funds:

This term represents the amount of balance the lender has to finance through the cost of funds r_c .

$$COF(t) = S(t)\bar{B}_c(t)r_c \tag{1.26}$$

Notice that the ALM unit will have to fund the behavioral expected balance, i.e the remaning balance after deducting prepayments (either full or partial) and defaults. ³

1.5.3 Equity Benefit (Capital Rebate) and Equity Capital Charge:

Since any loan has to be financed by both debt and capital, let's assume the fraction of capital the lender has to mantain for each lended dollar is α_t . To be consistent with the left and right side of the balance sheet, we need to take into account both the additional revenue of investing the amount of required capital (Equity Benefit) and also the equity charge that the shareholder demands.

$$EB(t) = \alpha_t S(t) \bar{B}(t) r_c \tag{1.27}$$

$$EC(t) = \alpha_t S(t) \bar{B}(t) r_e \tag{1.28}$$

1.5.4 Loss from Default:

So far, we have included the effect of the client's behavior on the interest income and outcome. In this term, we will also incorporate the loss of capital in which the lending entity incur when a client defaults.

$$EL(t) = p_d(t)LGD(t)S(t)\bar{B}(t)$$
(1.29)

Another valid alternative might be the one suggested by Phillips (2018): $COF(t) = S_c(t)\bar{B}_c(t)r_c$, where: $S_c(t) = \prod_{s=0}^{t} [1 - p_p(s) - (1 - LGD(s))p_d(s)]$

1.5.5 Fees and Servicing Costs:

The lending entity can get additional source of revenue or incur in costs that do not depend on the lended amount can be written as:

$$F(t) = fS(t) \tag{1.30}$$

$$SC(t) = \sigma S(t) \tag{1.31}$$

1.5.6 Recovery costs

$$C(t) = c \times p_d(t)S(t) \tag{1.32}$$

1.6 Incremental Profit Definition (CLV):

The net present value is given by:

$$NPV(x(t), r, T) = \sum_{t=1}^{T} \frac{x(t)}{(1+r)^t}$$
 (1.33)

Element	Notation	Calculation
Lending Interest	LI	$NVP(LI(t), r_d, T)$
Cost of Funds	COF	$NVP(COF(t), r_d, T)$
Equity benefit	EB	$NVP(EB(t), r_d, T)$
Fees	LI	$NVP(F(t), r_d, T)$
Ancillary profit	A	_
Origination cost	OC	_
Commission	COM	_
Servicing Costs	SC	$NVP(SC(t), r_d, T)$
Expected Loss	EL	$NVP(EL(t), r_d, T)$
Collection costs	C	$NVP(C(t), r_d, T)$
Equity charge	EC	$NVP(EC(t), r_d, T)$

Element	Notation	Calculation
Net Interest Income	NII	LI - COF + EB
Total Income	TI	NII + A + F
Net Income before tax	NIBT	TI - OC - COM - SC - LD - C
Net Income after tax	NIAT	$(1-\tau) \times NIBT$
Incremental profit	IP	NIAT - EC

1.6.1 Incremental Profit Function:

Define the incremental profit function as:

$$\pi(p) = IP(p, r_d, B, T, \theta) \tag{1.34}$$

Given definition (1.34) we can perform several types of computations. For example, to compute the IRR of a given rate p we set $IP(p, irr, B, T, \theta) = 0$. To compute the minimum rate that covers all costs (and risks) and yields a profitability of r_d we set $IP(r^{min}, r_d, B, T, \theta) = 0$. Notice that given this setup, charging the minimum rate does not mean the lending entity is not making money, it just means we are charging enough to reach a target profitability rate.

The PricingPy Python Library

2.1 Implementation in vectorial form

In the previous chapter we showed the full model and expressed in such a simple final equation that it can be stated directly in vectorial form in Python or any programatic language with support for matrix algebra.

$$T = \begin{bmatrix} T_1 & \dots & T_m \end{bmatrix} \tag{2.1}$$

$$r = \begin{bmatrix} r_1 & \dots & r_m \end{bmatrix} \tag{2.2}$$

$$disb = \begin{bmatrix} d_1 & \dots & d_m \end{bmatrix} \tag{2.3}$$

$$T^{mat} = \begin{bmatrix} 1 \\ 2 \\ \vdots \\ T^{max} \end{bmatrix}$$
 (2.4)

With the definitions above, \bar{B} can be directly computed using equation (1.16) and array broadcasting operators in numpy:

$$\bar{B} = disb * \frac{(1+r)^{T^{max}} - (1+r)^{T^{mat}-1}}{(1+r)^{T^{max}} - 1}$$
(2.5)

With this approach \bar{B} results of order $T^{max} \times m$ as expected.

$$\bar{B} = \begin{bmatrix} b_{1,1} & \dots & b_{1,m} \\ \vdots & \ddots & \\ b_{T^{max},1} & b_{T^{max},m} \end{bmatrix}$$

$$(2.6)$$

2.2 Class Diagram

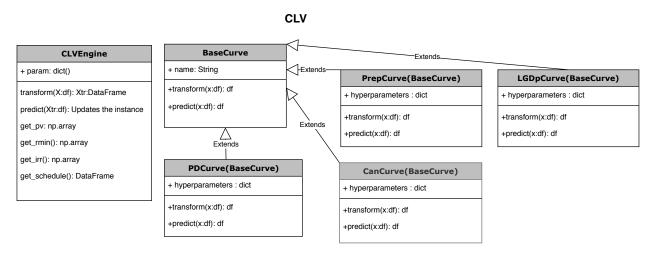


Figure 2.1: CLV Class Diagram

SKLearn Estimator ColumnTrasformer(List) + hyperparameter: hyperparameters + list : List +fit(x) + fit: + fit_transform: df, np.array Extends Transformer(Estimator) + hyperparameters : dict +transform(x): df, np.array +fit_transform(x): df, np.array Extends Pipeline(List) + list : List Predictor(Transformer) + fit: +predict(x): df, np.array + fit_transform: df, np.array

Figure 2.2: SKLearn Class Diagram

```
1 import pandas as pd
2 import numpy as np
3 import PyPricing as ppr
5 # Loading Data
6 filename = 'inputs.xlsx'
7 full_filename = Path('.').resolve() / 'data'/ filename
8 X = pd.read_excel(full_filename, sheet_name = 'inputs')
10 # Creating CLV engine to compute CLV
eng = ppr.CLVEngine()
12 Xtr = eng.transform(X)
eng = eng.predict(Xtr) # computes behavioral curves and update all
     behavioral curves
14
15 # Computations
16 PV = eng.get_pv()
RMIN = eng.get_rmin()
18 IRR = eng.get_irr()
19 SCHED = eng.get_schedule()
```

Listing 2.1: Python example

Willingness to Pay Modeling (WTP):

3.1 The Price-Response Function

Each price-response function specifies the demand that the lender would experience at each price, which will depend on:

- 1. Total number of clients interested in the loan
- 2. Number of applicant clients
- 3. The number of applicant the lender deems creditworthy and quotes a price.
- 4. Number of accepted applicants who would achieve a positive surplus from taking the loan from the lender at the offered price.
- 5. Number of accepted applicants who **take-up** the offered loan.

In most lending markets, the final price is not known to the client at the time she applies for the loan so we assume that the number of clients who apply for a loan is not influenced by the price.

$$d(p) = D\bar{F}(p) \tag{3.1}$$

d(p) is the number of the loans offered by a lender that would be taken up at the price p. D is the number of successful applicants for the loan, and $\overline{F}(p)$ is the take-up rate, which is defined as the fraction of successful applicants who will take up the loan at price p

$$\bar{F}(p) = \int_{p}^{\infty} f(w)dw \tag{3.2}$$

3.2 Segmented vs Join Estimation

For n segments, the segmented estimation assumes each segments has its own demand function so we need to estimate 2n parameters.

$$\bar{F}_i(p_i) = \frac{e^{a_i + b_i p_i}}{1 + e^{a_i + b_i p_i}} \tag{3.3}$$

For n segments, the join estimation assume we can estimate one single price response function that includes all explanatory variables within it (including price)

$$\bar{F}(p_i, a, b, \theta, x_i) = \frac{e^{a + bp_i + \theta^T x_i}}{1 + e^{a + bp_i + \theta^T x_i}}$$
(3.4)

Price optimization (CLV+WTP):

Consider the incremental profit function given by $\pi(p)$ in equation (1.34) and the take-up rate function given by $d(p) = D\bar{F}(p)$ in equation (3.1). Moreover, following section (3.2), lets consider we are interested in optimizing across N client segments.

4.1 Price optimization without constraints

$$p^* = \arg\max_{p} \sum_{i=1}^{N} D_i \overline{F}_i(p_i) \pi(p_i)$$
(4.1)

4.2 Price optimization with competing objectives: The efficient Frontier

$$p^* = \arg\max_{p} \sum_{i=1}^{N} D_i \bar{F}_i(p_i) \pi(p_i)$$
 (4.2)

s.t.

$$\sum_{i=1}^{N} D_i \bar{F}_i(p_i) = q \tag{4.3}$$

Survival models

5.0.1 Standard survival setup

Let T be a positive random variable in 1, 2, 3, ...

$$S(t) = P(T > t) \tag{5.1}$$

$$F(t) = 1 - S(t) = 1 - P(T > t) = P(T \le t)$$
(5.2)

5.0.2 Survival setup in presence of competing risks

We define the cumulative incidence function as:

$$CIF_k(t) = P(T \le t, D = k) \tag{5.3}$$

$$= \sum_{k} P(T \le t, D = k) = P(T \le t)$$
 (5.4)

In order to see what is the relationship between the CIF function and the usual conditional probability of default (death) we state the definition of conditional probability and use the fact that the event $T = t + 1 \land T > t$ is equal to T = t + 1 standalone.

$$p_k(t+1) = p(T=t+1, D=k/T > t) = \frac{P(T=t+1, D=k)}{P(T>t)}$$
(5.5)

$$= \frac{P(T \le t+1, D=k) - P(T \le t, D=k)}{1 - \sum_{k} P(T \le t, D=k)}$$
(5.6)

As an example lets consider that D=1 represents default and D=2 represents prepayment then the conditional probabilities of default and prepayment are given by:

$$p_d(t+1) = \frac{CIF_d(t+1) - CIF_d(t)}{1 - CIF_d(t) - CIF_p(t)}$$
(5.7)