3rd Seminar on Data Science and Analytics Bangalore July 27th, 2019

Use cases of Financial Data Science Techniques in Retail Analytics

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Survival Models



Two kinds of Survival Models

Event Model:

modeling probability of surviving beyond a time threshold

> Example: Attrition Probability

Popular technique: Cox Regression

Waiting time model:

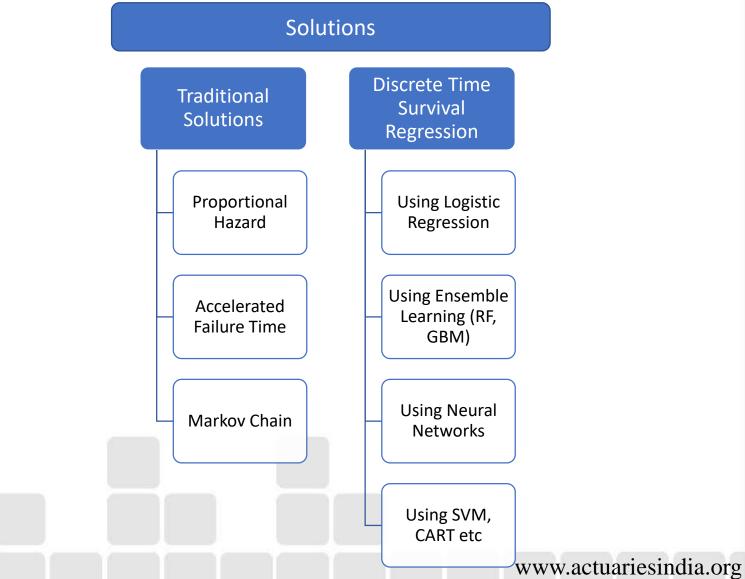
modeling waiting time till an event

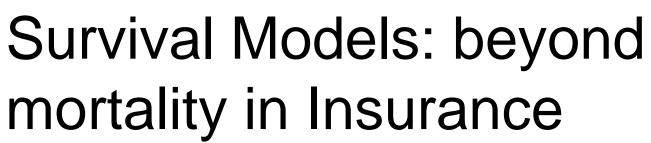
Example: Attrition Timing

Popular Technique: AFT

Survival Models

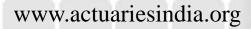








- Lapse
- Survival of reserves in non-life



Customer Lifetime Value & Attrition



In insurance:

• $CLTV = \sum_{t=1}^{\infty} e^{-rt} \times P(A \ge t) \times [P(S \ge t) \times I_t + P(S = t) \times D_t]$

S= Remaining lifetime after time 0

A = Time to attrition

 I_t = Net income at time t if survives time t (usually positive)

 D_t = Net income due to death at time t (usually negative)

Insurance: Mortality and Lapse both important Long tenure Traceable customer Death is usually reported

Retail: Lapse contains mortality Short Tenure (not really a lifetime) Not 100% traceable customers Death is untraceable

Customer Value in Retail



- Structural CLTV approach:
 - Survival model and a Money model
- Distribution based CLTV: Buy Till You Die (BTYD)
 - Two distributions calibrated from customer's own data
 - Probability of Being Alive
 - Probability of a purchase
 - Pareto/NBD, Beta-Geometric(BG)/NBD
- RFM or Recency Frequency Money
 - Customer Lifetime Value

Survival Models: beyond Cox & AFT



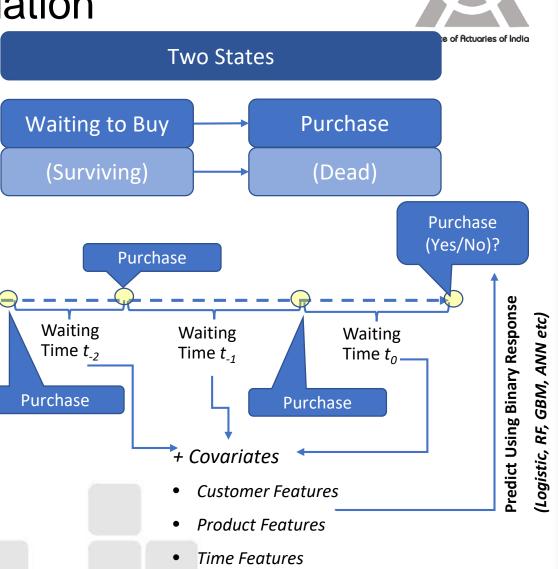
• Discrete time survival regression:

- GLM based (Logistic/Poisson)
- Ensemble Learning (Random Forest/ GBM)
- Others (SVM/ANN)
- Advantages:
 - No nonparametric estimation for baseline hazard
 - Handling of high dimensionality (Regularization/Ensemble)

Survival Models for Product Recommendation

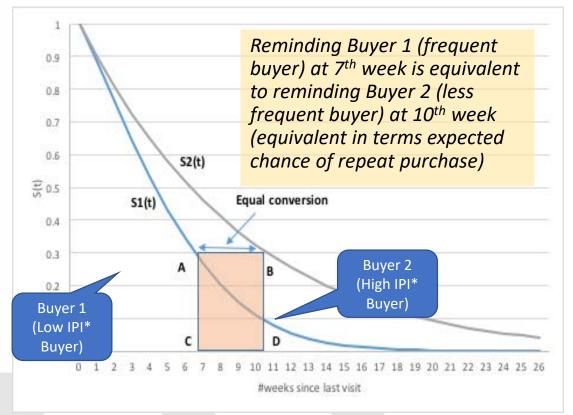
Product recommendation:

- Waiting time: Time between two purchase for repeating buyers
- A purchase is an event which is a function of how long the customer is waiting in a non-purchase state and other covariates (product features, customer features and time features)



Survival Models for Product Recommendation

Product recommendation as a waiting process until the customer's natural comeback time overshoots



*IPI= Inter Purchase Interval; S(t)= P[waiting time > t] S(t)= Survival Probability i.e. Pr(Wait for customers come back continues beyond t) ::: Strictly non-increasing function of t www.actuariesindia.org

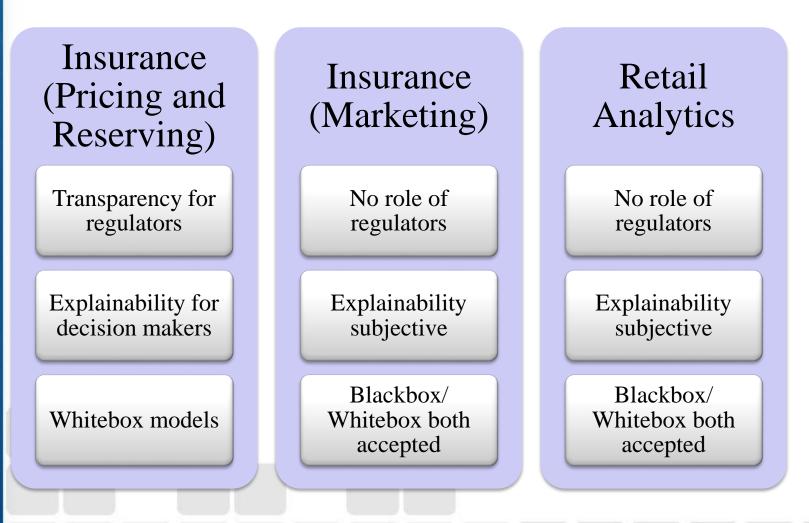
Survival Models: other use

- Inventory management:
 - Next demand generation as a function of time since previous demand generation

- Next supply as a function of time since previous supply
- Queuing theory

The way we model: differences

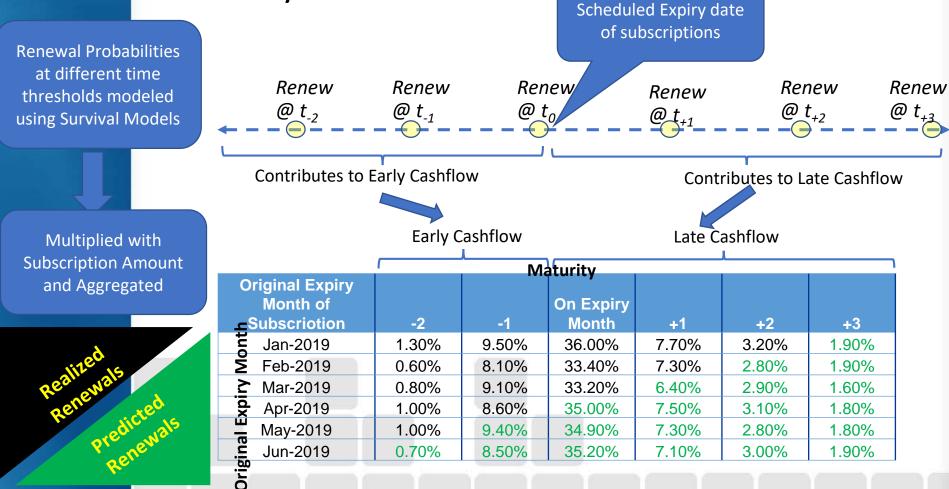






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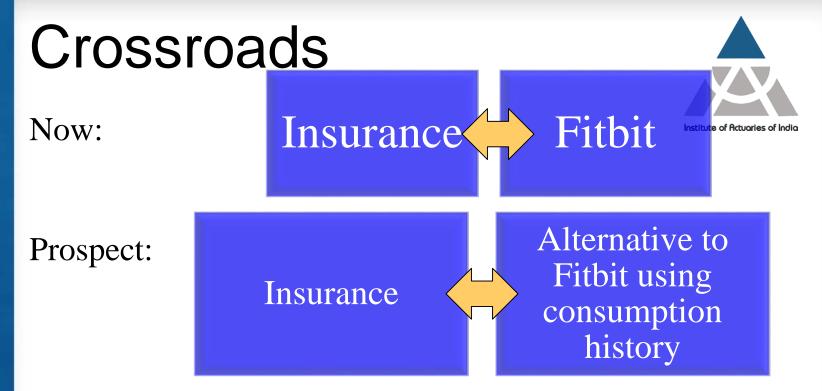
 An annual subscription based business offering flexibility in renewal date



Cashflow Models: main differences



- Life Insurance (usual scenario):
 - Positive Cashflow on survival
 - Negative cashflow on death/maturity
 - Traceable death
- Pension:
 - Negative Cashflow on survival
 - No cashflow on death
 - Traceable death
- Retail Subscription:
 - Positive cashflow on survival
 - No cashflow on death
 - Untraceable death



Challenges:

- Personal Data
- Adverse selection and bias on both side through voluntary participation

Opportunities:

- Profiling at geographic level