

ABA Recitation 1

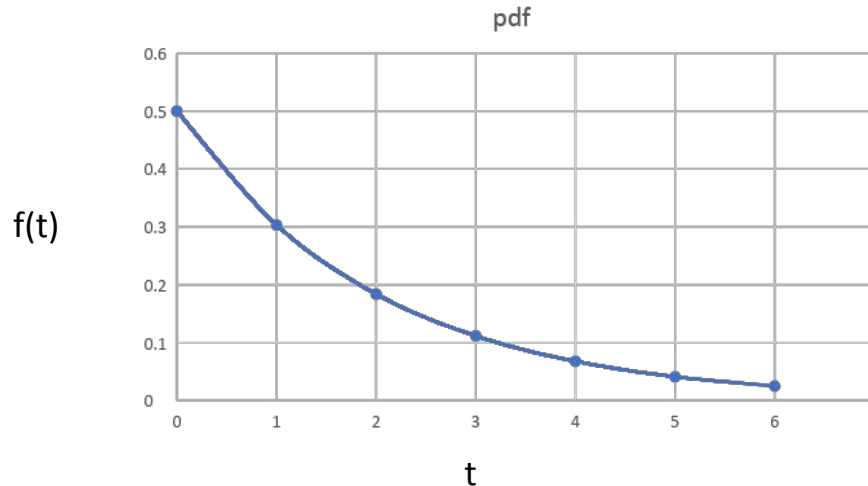
Spring 2024
01/26/24

Agenda

- Quiz 1 is out, due Jan 29 11:59PM
- Hazard Functions:
 - PDF and CDF
 - Hazard function example (constant Hazard)
 - From hazard to CDF $F(t)$ and PDF $f(t)$
 - Empirical Hazard function (google sheet)
- OLS example in Python

Probability Density Function (PDF) and CDF (Cumulative Density Function)

- What is pdf?
- For Exponential pdf $f(t) = \lambda e^{-\lambda t}$
- $f(t)$ outlines how data is distributed with time t , given the parameter λ . The following figure shows how data is distributed.



F(t)

- Cumulative distribution is

- $F(t) = \int_0^t f(x) dx$

- $F(t) = \int_0^t \lambda e^{-\lambda x} dx$

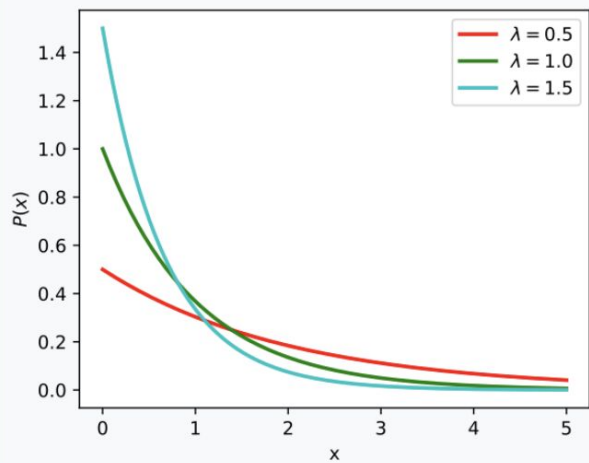
- Or $F(t) = \lambda \int_0^t e^{-\lambda x} dx$

- Or $\lambda \left(-\frac{1}{\lambda} [e^{-\lambda x}]_0^t \right)$

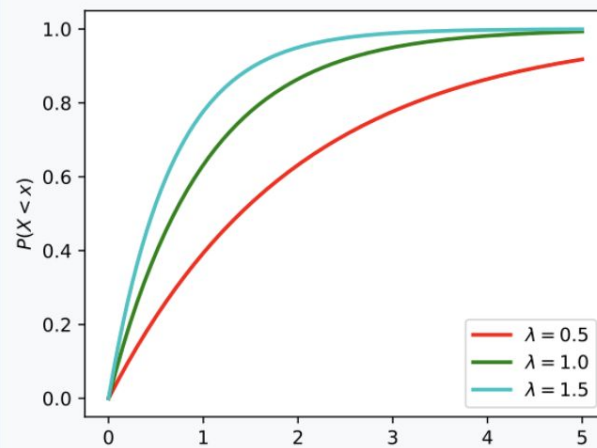
- Or $(1 - e^{-\lambda t})$

Exponential

Probability density function



Cumulative distribution function



Hazard Functions

- For analysing **timing question** (when will something happen?)
- Hazard:
 - What is the probability that the event will happen *right now*, given that it has *not yet happened*?
 - This is called the *hazard rate*.

Starting Point for Hazard Analysis

- Exponential function is a commonly used function.

$$\begin{aligned}h(t) &= \frac{f(t)}{1 - F(t)} \\&= \frac{\lambda e^{-\lambda t}}{e^{-\lambda t}} \\&= \lambda\end{aligned}$$

- Notice that exponential implies a “constant” hazard rate.
 - This is why exponential is called a “memory-less” distribution because hazard is not a function of time.
- $S(T) = 1 - F(T)$ is the survivor function.

Hazard Rate and Distribution Function

- Remember that Hazard functions have one-on-one mapping with the distribution function. If you are defining one, you are automatically defining the other.

$$F(t) = 1 - \exp\left(-\int_0^t h(u) du\right)$$

- Sometimes it is much easier to define Hazard function first.
- By defining hazard rate, one has timing process to work with.

Derivation: from $h(t)$ to $F(t)$ to $f(t)$ for exponential

- In exponential hazard $= \lambda$

Derivation: from $h(t)$ to $F(t)$ to $f(t)$ for exponential

- In exponential hazard $= \lambda$
- We know that

$$F(t) = 1 - \exp\left(-\int_0^t h(u) du\right)$$

- Since $h()$ is a constant, integration of constant is simply $h() \cdot u$ or

$$\text{So } F(t) = 1 - \exp\left(-\int_0^t \lambda du\right)$$

$$\text{Or } F(t) = 1 - \exp(-\lambda t)$$

Distributions

- Commonly used distribution for survival analysis?

Exponential pdf

$$f(t) = \lambda e^{-\lambda t}$$

Exponential cdf

$$F(t) = 1 - e^{-\lambda t}$$

Hazard

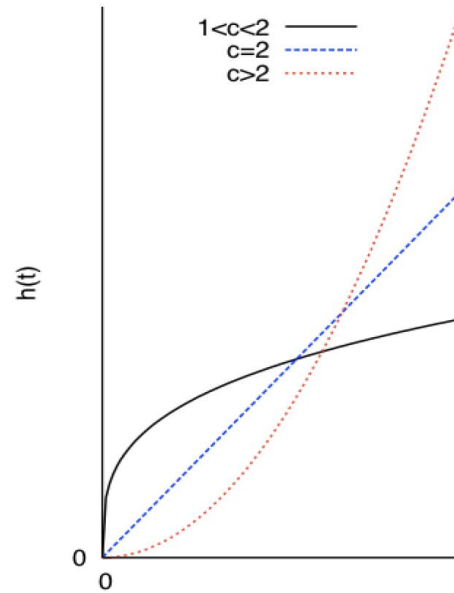
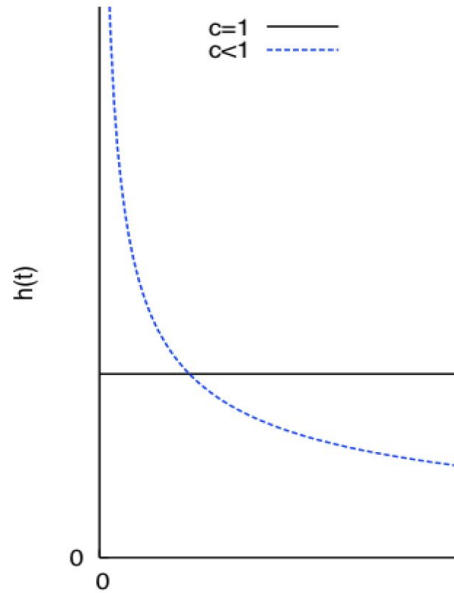
$$h(t) = \frac{f(t)}{1 - F(t)} = \frac{\lambda e^{-\lambda t}}{e^{-\lambda t}} = \lambda$$

Constant hazard. That is the hazard rate is not a function of time.

Weibull Distribution

- A flexible distribution that can represent increasing or decreasing hazard rates readily. This allows for λ to change with time
 - $F(t) = 1 - e^{-\lambda t^c}$
 - $f(t) = \lambda c t^{c-1} e^{-\lambda t^c}$
 - $h(t) = \frac{f(t)}{1-F(t)} = \lambda c t^{c-1}$
- The hazard is function of time. That means depending on c , the hazard can be increasing or decreasing. For $c=1$, this boils down to exponential

Weibull Hazard



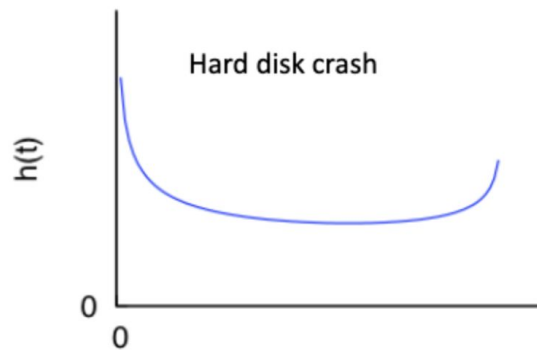
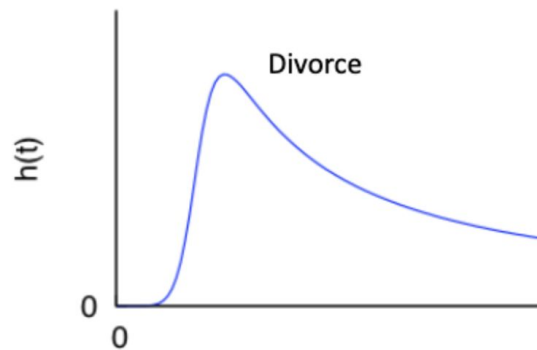
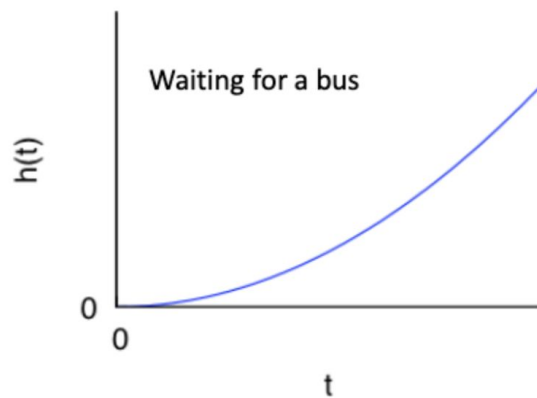
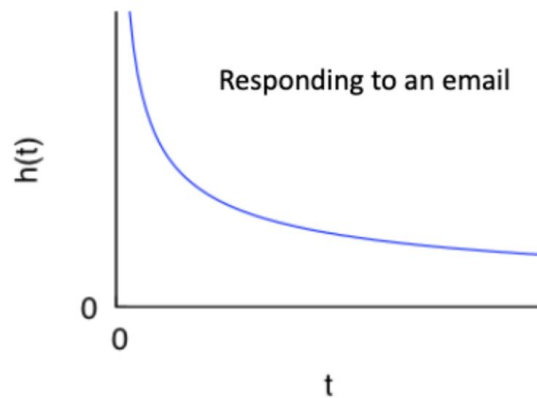
- Decreasing hazard rate (negative duration dependence) when $c < 1$
- Increasing hazard rate (positive duration dependence) when $c > 1$

Empirical Hazard Example

See google sheet:

https://docs.google.com/spreadsheets/d/1feqc_Sry5nKtmHXAkZkFGTA3TRANq9GNJMmOV0aS8k/edit?usp=sharing

Some shapes of hazard rate functions



OLS in Python

See jupyter notebook

Note on libraries for next homeworks

install lifelines to use for KM in python

```
#conda install -c conda-forge lifelines
```