

# 24) Survival Function and Hazard Rates

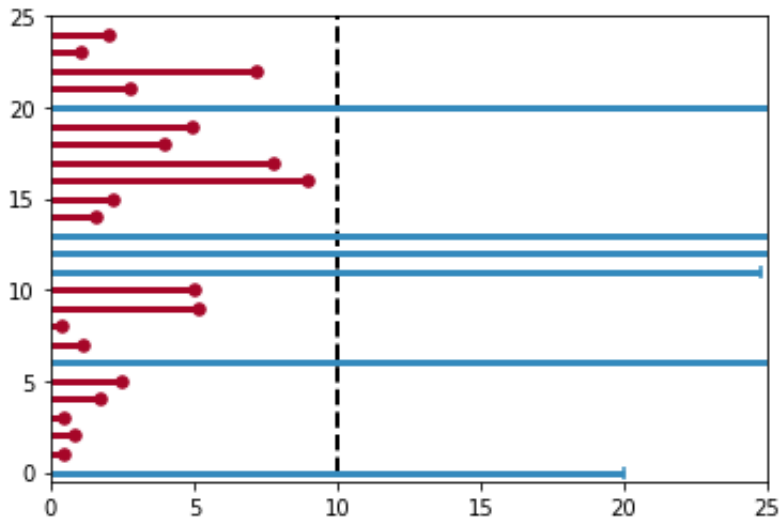
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Tables, Graphics, and Figures from  
**<https://lifelines.readthedocs.io/>**

Introduction to Survival Analysis

# Simulated Data



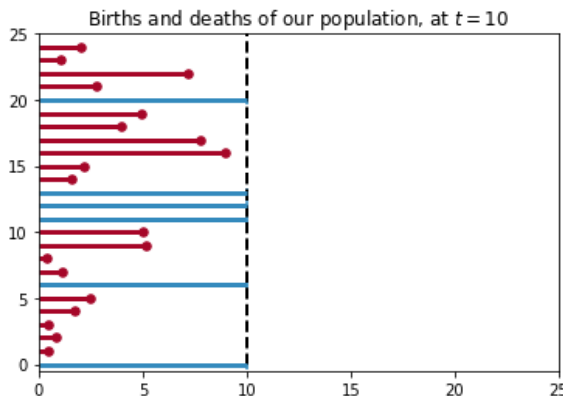
# Generated Data

```
from lifelines.plotting import plot_lifetimes
from numpy.random import uniform, exponential
import numpy as np
import matplotlib.pyplot as plt
```

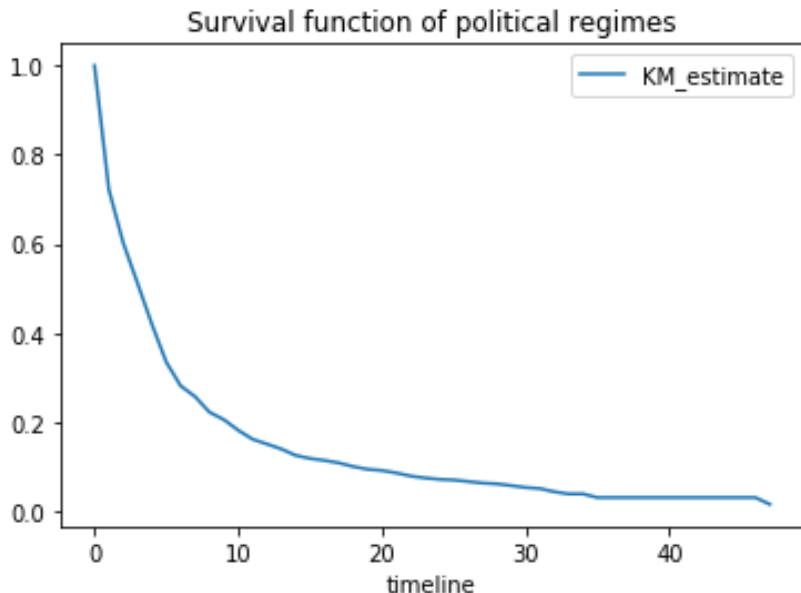
```
N = 25
current_time = 10
actual_lifetimes = np.array([[exponential(12),
    exponential(2)])[uniform() < 0.5] for i in range(N)])
observed_lifetimes = np.minimum(actual_lifetimes,
    current_time)
observed = actual_lifetimes < current_time
```

# Observed Lifetimes at Time 10

```
plt.xlim(0, 25)
plt.vlines(10, 0, 30, lw=2, linestyle='--')
plt.xlabel("time")
plt.title("Births and deaths of our population, at $t=10$")
plot_lifetimes(observed_lifetimes, event_observed=observed)
```



# Survival Function of Political Regimes



# Survival Function

Prob. of surviving past time  $t$

Prob. death event has not occurred yet at time  $t$

$$S(t) = \Pr(T > t)$$

$$0 \leq S(t) \leq 1$$

$F(t) = 1 - S(t)$ , where  $F(t)$  is the *CDF* of  $T$

$S(t)$  is non-increasing function of  $t$

# Kaplan-Meier Estimate of Survival Function

$$\hat{S}(t) = \prod_{t_j < t} \frac{n_j - d_j}{n_j}$$

$d_j$  : # of death events at time  $t$

$n_j$  : # of subjects at risk of death just prior to time  $t$



# Import Data

```
import pandas as pd
from lifelines.datasets import load_dd
data = load_dd()
data.sample(6)
```

|      |   | leaderspell     | reg | democracy     |
|------|---|-----------------|-----|---------------|
| 369  | Nikica Valentic.Croatia.1993.1994.                | Mixed           | Dem | Democracy     |
| 1645 | Suleyman Demirel.Turkey.1975.1977.                | Parliamentar... |     | Democracy     |
| 827  | Silvio Berlusconi.Italy.2008.2008.                | Parliamentar... |     | Democracy     |
| 1349 | Armando Vaz d'Almeida.Sao Tome and Principe.19... |                 |     | Democracy     |
| 1154 | Ali Saibou.Niger.1987.1992.                       | Military Dict   |     | Non-democracy |

|      | regime            | start_year | duration | observed |
|------|-------------------|------------|----------|----------|
| 369  | Mixed Dem         | 1993       | 2        | 1        |
| 1645 | Parliamentary Dem | 1975       | 3        | 1        |
| 827  | Parliamentary Dem | 2008       | 1        | 0        |
| 1349 | Mixed Dem         | 1995       | 1        | 1        |
| 1154 | Military Dict     | 1987       | 6        | 1        |

# Kaplan Meier Fitter

```
from lifelines import KaplanMeierFitter  
kmf = KaplanMeierFitter()
```

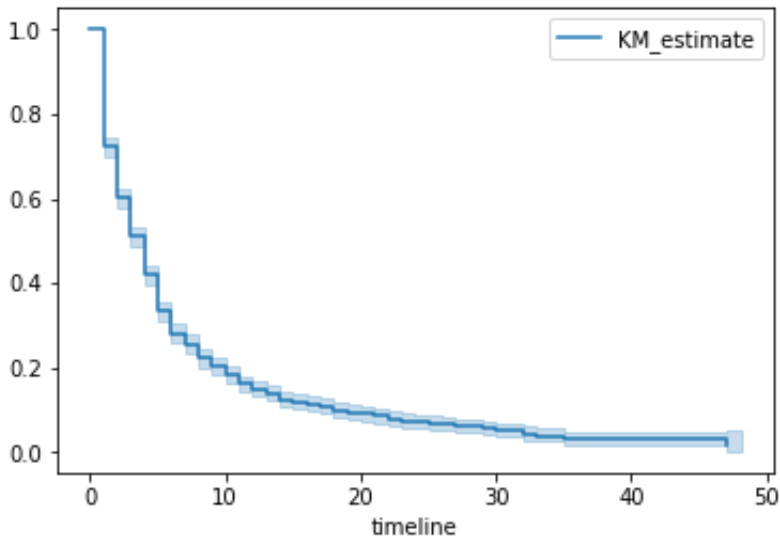
```
T = data["duration"]  
E = data["observed"]
```

```
kmf.fit(T, event_observed=E)
```

```
kmf.median_
```

4

# kmf.plot()



# Democratic vs Non-Democratic Regimes

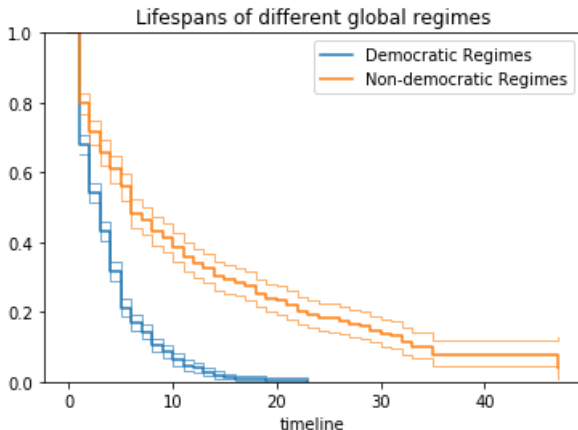
```
ax = plt.subplot(111)
dem = (data["democracy"] == "Democracy")

t = np.linspace(0, 50, 51)
kmf.fit(T[dem], event_observed=E[dem], timeline=t,
        label="Democratic Regimes")
ax = kmf.plot(ax=ax)
print("Median survival time of democratic:", kmf.median_)

kmf.fit(T[~dem], event_observed=E[~dem], timeline=t,
        label="Non-democratic Regimes")
ax = kmf.plot(ax=ax)
print("Median survival time of non-democratic:", kmf.median_)

plt.ylim(0,1)
plt.title("Lifespans of different global regimes");
```

# Lifespans of Different Global Regimes



Median of democratic: **3.0**

Median non-democratic: **6.0**

# Logrank Test (Mantel-Cox Test)

```
from lifelines.statistics import logrank_test
results = logrank_test(T[dem], T[~dem],
                        E[dem], E[~dem], alpha=.99)
results.print_summary()
```

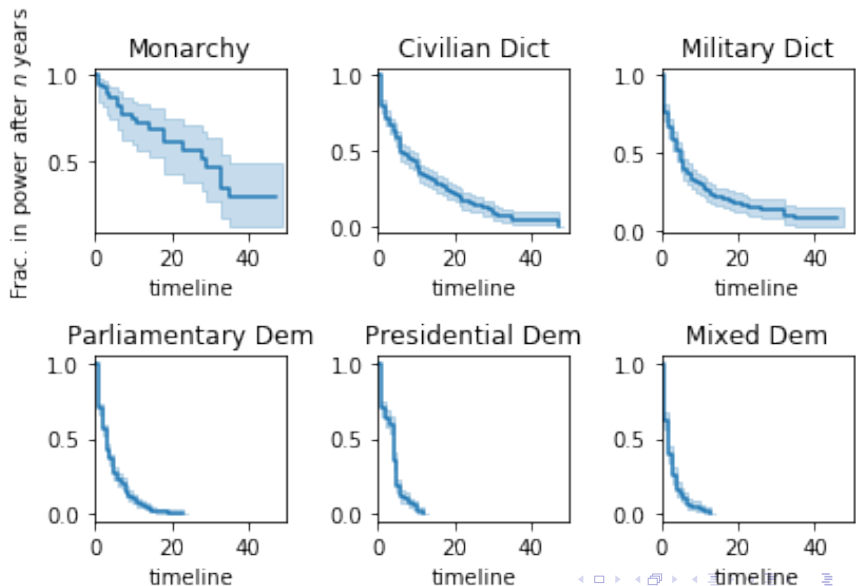
```
t_0=-1, alpha=0.99, null_distribution=chi squared, df=1
```

```
test_statistic      p
      260.4695 0.0000 ***
```

---

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

# Lifespans of Regime Types



# Hazard Curve

Prob. of death event occurring at time  $t$ , given that the death event has not occurred until time  $t$

$$\lambda(t) = \lim_{\Delta t \rightarrow 0} \frac{Pr(t \leq T < t + \Delta t | T \geq t)}{\Delta t}$$

$$\lambda(t) = \frac{f(t)}{S(t)} = \frac{-S'(t)}{S(t)} = \frac{-d \ln(S(t))}{dt}$$

$$S(t) = \exp\left(-\int_0^t \lambda(z) dz\right)$$



# Estimating Hazard Rates by Nelson-Aalen Estimator

$$\Lambda(t) = \int_0^t \lambda(z) dz = -\ln S(t)$$

$$\hat{\Lambda}(t) = \sum_{t_j \leq t} \frac{d_j}{n_j}$$

$d_j$  : # of deaths at time  $t_j$

$n_j$  : # of susceptible individuals

```
from lifelines import NelsonAalenFitter
```

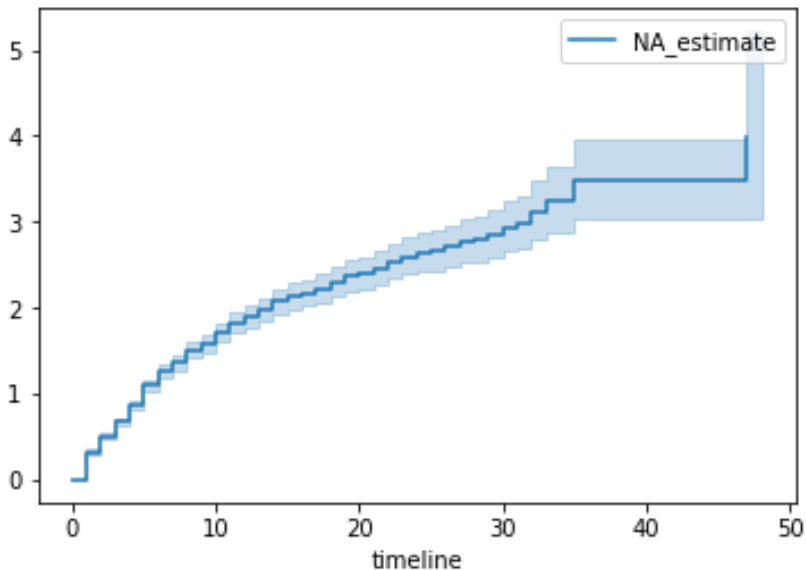
```
naf = NelsonAalenFitter()
```

```
naf.fit(T,event_observed=E)
```

```
print(naf.cumulative_hazard_.head())
```

| timeline | NA_estimate |
|----------|-------------|
| 0.0      | 0.000000    |
| 1.0      | 0.325912    |
| 2.0      | 0.507356    |
| 3.0      | 0.671251    |
| 4.0      | 0.869867    |

# naf.plot()



# Cumulative Hazard Function

```
naf.fit(T[dem], event_observed=E[dem], label="Democratic Regimes")  
ax = naf.plot(loc=slice(0, 20))  
naf.fit(T[~dem], event_observed=E[~dem], label="Non-democratic Regimes")  
naf.plot(ax=ax, loc=slice(0, 20))
```

