

# Modelo Cox - Análisis de Recesión

Juan Leal

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```
import pandas as pd
import numpy as np
import statsmodels.api as sm
from lifelines import ExponentialFitter, WeibullFitter, CoxPHFitter
```

```
%store -r df_cox
```

```
df_cox = df_cox[df_cox['duration']>0]
df_cox
```

|    | id          | duration | event | GDP       | Personal_Income | TasaDesempleo |
|----|-------------|----------|-------|-----------|-----------------|---------------|
| 0  | Alabama     | 10       | 0     | 12.196037 | 11.964237       | 8.009091      |
| 1  | Alaska      | 10       | 0     | 10.863448 | 10.404198       | 7.351515      |
| 2  | Arizona     | 10       | 0     | 12.574947 | 12.289487       | 7.875758      |
| 3  | Arkansas    | 1        | 1     | 11.629167 | 11.427889       | 6.681818      |
| 4  | California  | 10       | 0     | 14.582742 | 14.266541       | 9.578788      |
| 5  | Colorado    | 9        | 0     | 12.538412 | 12.228448       | 6.684848      |
| 6  | Connecticut | 9        | 0     | 12.509942 | 12.270401       | 7.203030      |
| 7  | Delaware    | 9        | 0     | 11.120425 | 10.504921       | 6.612121      |
| 8  | Florida     | 10       | 0     | 13.672976 | 13.472931       | 8.463636      |
| 9  | Georgia     | 10       | 0     | 13.068082 | 12.713747       | 8.266667      |
| 10 | Hawaii      | 10       | 1     | 11.233874 | 10.924578       | 5.345455      |
| 11 | Idaho       | 10       | 0     | 11.043447 | 10.806564       | 7.151515      |
| 12 | Illinois    | 10       | 0     | 13.530172 | 13.195753       | 8.466667      |
| 13 | Indiana     | 9        | 0     | 12.655854 | 12.319331       | 8.187879      |

|    | id             | duration | event | GDP       | Personal_Income | TasaDesempleo |
|----|----------------|----------|-------|-----------|-----------------|---------------|
| 14 | Iowa           | 9        | 0     | 11.979158 | 11.646409       | 5.360606      |
| 15 | Kansas         | 7        | 1     | 11.877897 | 11.630400       | 5.890909      |
| 16 | Kentucky       | 9        | 0     | 12.117988 | 11.849883       | 8.651515      |
| 17 | Louisiana      | 8        | 0     | 12.393479 | 12.029710       | 5.772727      |
| 18 | Maine          | 9        | 0     | 10.993381 | 10.804696       | 7.036364      |
| 19 | Maryland       | 8        | 0     | 12.733998 | 12.536470       | 5.896970      |
| 20 | Massachusetts  | 9        | 0     | 13.010134 | 12.716271       | 6.824242      |
| 21 | Michigan       | 9        | 0     | 12.990805 | 12.758253       | 10.996970     |
| 22 | Minnesota      | 10       | 0     | 12.622804 | 12.304616       | 6.603030      |
| 23 | Mississippi    | 8        | 0     | 11.584078 | 11.408196       | 8.206061      |
| 24 | Missouri       | 10       | 0     | 12.591244 | 12.291466       | 7.763636      |
| 25 | Montana        | 10       | 0     | 10.638472 | 10.445304       | 6.024242      |
| 26 | Nebraska       | 1        | 1     | 11.518610 | 11.186013       | 4.012121      |
| 27 | Nevada         | 10       | 0     | 11.894895 | 11.514505       | 9.393939      |
| 28 | New Hampshire  | 9        | 0     | 11.173738 | 11.000698       | 5.027273      |
| 29 | New Jersey     | 9        | 0     | 13.235134 | 13.001712       | 7.366667      |
| 30 | New Mexico     | 9        | 0     | 11.405901 | 11.115575       | 5.896970      |
| 31 | New York       | 9        | 0     | 14.128133 | 13.729350       | 7.048485      |
| 32 | North Carolina | 9        | 0     | 13.086100 | 12.746720       | 8.548485      |
| 33 | North Dakota   | 3        | 1     | 10.483193 | 10.203878       | 3.530303      |
| 34 | Ohio           | 9        | 0     | 13.246421 | 12.937403       | 8.551515      |
| 35 | Oklahoma       | 7        | 1     | 11.966133 | 11.829367       | 5.239394      |
| 36 | Oregon         | 9        | 0     | 12.106556 | 11.824306       | 8.600000      |
| 37 | Pennsylvania   | 9        | 0     | 13.394662 | 13.158647       | 6.818182      |
| 38 | Rhode Island   | 10       | 0     | 10.914361 | 10.679802       | 9.506061      |
| 39 | South Carolina | 9        | 0     | 12.148602 | 11.907656       | 9.072727      |
| 40 | South Dakota   | 1        | 1     | 10.683103 | 10.382696       | 3.957576      |
| 41 | Tennessee      | 10       | 1     | 12.588919 | 12.295014       | 8.436364      |
| 42 | Texas          | 7        | 0     | 14.072110 | 13.758478       | 6.415152      |
| 43 | Utah           | 9        | 0     | 11.794488 | 11.385811       | 5.745455      |
| 44 | Vermont        | 9        | 0     | 10.307443 | 10.138916       | 5.539394      |
| 45 | Virginia       | 9        | 0     | 13.045591 | 12.774823       | 5.521212      |
| 46 | Washington     | 9        | 0     | 12.893091 | 12.553836       | 7.463636      |
| 47 | West Virginia  | 2        | 1     | 11.167021 | 10.971535       | 6.506061      |
| 48 | Wisconsin      | 1        | 1     | 12.553049 | 12.289708       | 7.054545      |
| 49 | Wyoming        | 9        | 0     | 10.610985 | 10.158213       | 4.957576      |

```
#Modelo Exponencial
exp_model = ExponentialFitter()
exp_model.fit(df_cox['duration'], event_observed=df_cox['event'])
```

```
summary = exp_model.summary
print(summary)
```

|         | coef | se(coef)  | coef lower 95% | coef upper 95% | cmp to | z \      |
|---------|------|-----------|----------------|----------------|--------|----------|
| lambda_ | 41.1 | 12.996961 | 15.626424      | 66.573576      | 0.0    | 3.162278 |

  

|         | p        | -log2(p) |
|---------|----------|----------|
| lambda_ | 0.001565 | 9.319251 |

```
# Weibull
weibull_model = WeibullFitter()
weibull_model.fit(df_cox['duration'], event_observed=df_cox['event'])
summary = weibull_model.summary
print(summary)
```

|         | coef      | se(coef)  | coef lower 95% | coef upper 95% | cmp to | \ |
|---------|-----------|-----------|----------------|----------------|--------|---|
| lambda_ | 51.013950 | 33.072391 | -13.806746     | 115.834645     | 1.0    |   |
| rho_    | 0.877066  | 0.268032  | 0.351733       | 1.402399       | 1.0    |   |

  

|         | z         | p        | -log2(p) |
|---------|-----------|----------|----------|
| lambda_ | 1.512257  | 0.130469 | 2.938226 |
| rho_    | -0.458655 | 0.646482 | 0.629318 |

```
#CoX
cox = CoxPHFitter()
cox.fit(df_cox, duration_col='duration', event_col='event', formula="GDP")
summary = cox.summary
print(summary)
```

|           | coef      | exp(coef) | se(coef) | coef lower 95% | coef upper 95% | \ |
|-----------|-----------|-----------|----------|----------------|----------------|---|
| covariate |           |           |          |                |                |   |
| GDP       | -0.675568 | 0.508867  | 0.342842 | -1.347525      | -0.003611      |   |

  

|           | exp(coef) | lower 95% | exp(coef) | upper 95% | cmp to | z \       |
|-----------|-----------|-----------|-----------|-----------|--------|-----------|
| covariate |           |           |           |           |        |           |
| GDP       | 0.259883  |           | 0.996396  |           | 0.0    | -1.970495 |

  

|           | p        | -log2(p) |
|-----------|----------|----------|
| covariate |          |          |
| GDP       | 0.048782 | 4.357518 |

```
#CoX
cox = CoxPHFitter()
cox.fit(df_cox, duration_col='duration', event_col='event', formula="GDP+Personal_Income")
summary = cox.summary
print(summary)
```

|                 | coef      | exp(coef) | se(coef) | coef lower 95% \ |
|-----------------|-----------|-----------|----------|------------------|
| covariate       |           |           |          |                  |
| GDP             | -4.597705 | 0.010075  | 3.932998 | -12.306239       |
| Personal_Income | 3.976648  | 53.337967 | 3.997968 | -3.859224        |

|                 | coef      | upper 95% | exp(coef) | lower 95% | exp(coef)     | upper 95% | \ |
|-----------------|-----------|-----------|-----------|-----------|---------------|-----------|---|
| covariate       |           |           |           |           |               |           |   |
| GDP             | 3.110830  |           | 0.000005  |           | 22.439662     |           |   |
| Personal_Income | 11.812521 |           | 0.021084  |           | 134931.275090 |           |   |

|                 | cmp to |           | z        | p        | -log2(p) |
|-----------------|--------|-----------|----------|----------|----------|
| covariate       |        |           |          |          |          |
| GDP             | 0.0    | -1.169008 | 0.242401 | 2.044535 |          |
| Personal_Income | 0.0    | 0.994667  | 0.319898 | 1.644316 |          |

```
#CoX
cox = CoxPHFitter()
cox.fit(df_cox, duration_col='duration', event_col='event', formula="GDP+Personal_Income+TasaDesempleo")
summary = cox.summary
print(summary)
```

|                 | coef      | exp(coef)  | se(coef) | coef lower 95% \ |
|-----------------|-----------|------------|----------|------------------|
| covariate       |           |            |          |                  |
| GDP             | -4.763167 | 0.008539   | 4.160908 | -12.918397       |
| Personal_Income | 4.816902  | 123.581643 | 4.218673 | -3.451546        |
| TasaDesempleo   | -0.910240 | 0.402428   | 0.341099 | -1.578782        |

|                 | coef      | upper 95% | exp(coef) | lower 95% | exp(coef)     | upper 95% | \ |
|-----------------|-----------|-----------|-----------|-----------|---------------|-----------|---|
| covariate       |           |           |           |           |               |           |   |
| GDP             | 3.392063  |           | 0.000002  |           | 29.727202     |           |   |
| Personal_Income | 13.085350 |           | 0.031697  |           | 481831.489049 |           |   |
| TasaDesempleo   | -0.241698 |           | 0.206226  |           | 0.785293      |           |   |

|           | cmp to |  | z | p | -log2(p) |
|-----------|--------|--|---|---|----------|
| covariate |        |  |   |   |          |

|                 |     |           |          |          |
|-----------------|-----|-----------|----------|----------|
| GDP             | 0.0 | -1.144742 | 0.252316 | 1.986697 |
| Personal_Income | 0.0 | 1.141805  | 0.253535 | 1.979743 |
| TasaDesempleo   | 0.0 | -2.668550 | 0.007618 | 7.036385 |

```
#CoX
cox = CoxPHFitter()
cox.fit(df_cox, duration_col='duration', event_col='event', formula="GDP+TasaDesempleo")
summary = cox.summary
print(summary)
```

|               | coef      | exp(coef) | se(coef) | coef lower 95% | coef upper 95% | \ |
|---------------|-----------|-----------|----------|----------------|----------------|---|
| covariate     |           |           |          |                |                |   |
| GDP           | -0.002251 | 0.997751  | 0.416038 | -0.817671      | 0.813169       |   |
| TasaDesempleo | -0.884599 | 0.412880  | 0.332099 | -1.535501      | -0.233696      |   |

|               | exp(coef) | lower 95% | exp(coef) | upper 95% | cmp to | z         | \ |
|---------------|-----------|-----------|-----------|-----------|--------|-----------|---|
| covariate     |           |           |           |           |        |           |   |
| GDP           |           | 0.441458  |           | 2.255042  | 0.0    | -0.005411 |   |
| TasaDesempleo |           | 0.215348  |           | 0.791602  | 0.0    | -2.663658 |   |

|               | p        | -log2(p) |
|---------------|----------|----------|
| covariate     |          |          |
| GDP           | 0.995682 | 0.006242 |
| TasaDesempleo | 0.007730 | 7.015389 |