# JP Morgan Startup Survival Rates The Continuous-Time Survival Rate Curve 

Gary Schurman MBE, CFA

March 2023

The startup survival rates from Part I are in discrete-time (annual increments). In this white paper we will map the JP Morgan survival rates (the corresponding failure rate is equal to one minus the survival rate) to a continuous-time curve. To that end we will work through the following hypothetical problem from Part I...

## Our Hypothetical Problem

We are given the following startup survival and failure rates by birth year...
Source: https://www.jpmorganchase.com/institute/research/small-business/small-business-dashboard/longevity

Table 1: JP Morgan Data

| Birth <br> Year | Survival <br> Rate | Failure <br> Rate |
| :---: | ---: | ---: |
| 0 | 1.0000 | - |
| 1 | 0.8010 | 0.1990 |
| 2 | 0.6870 | 0.1140 |
| 3 | 0.6160 | 0.0710 |
| 4 | 0.5600 | 0.0560 |
| 5 | 0.4880 | 0.0720 |
| 6 | 0.4470 | 0.0410 |
| 7 | 0.4160 | 0.0310 |
| 8 | 0.3800 | 0.0360 |
| 9 | 0.3550 | 0.0250 |
| 10 | 0.3380 | 0.0170 |
| 11 | 0.3160 | 0.0220 |
| 12 | 0.2990 | 0.0170 |
| 13 | 0.2840 | 0.0150 |
| 14 | 0.2660 | 0.0180 |
| 15 | 0.2510 | 0.0150 |
| 16 | 0.2400 | 0.0110 |
| 17 | 0.2300 | 0.0100 |
| 18 | 0.2220 | 0.0080 |
| 19 | 0.2130 | 0.0090 |
| 20 | 0.2040 | 0.0090 |
| Total | - | 0.7960 |

Table 2: Survival And Failure Rate Graph By Birth Year In Discrete-Time


Note that in the graph above survival rates are approximately Exponentially-distributed (i.e. the first derivative is negative (survival rates decrease over time) and the second derivative is positive (survival rates decrease over time but at a decreasing rate). The Exponential distribution is a one parameter probability distribution.

Question: Given that ABC Company has survived to year 3.25 , what is the probability that ABC company will survive until year 7.75?

## Mapping Survival Rates to an Exponential Distribution

We will define the function $S(t)$ to be the survival function at time $t$, which is the probability that a startup company survives over the time interval $[0, t]$ where time zero is the time in years that the company began operations. If the survival rate is an Exponentially-distributed random variable then the equation for the expected survival rate at time $t$ is... [1]

$$
\begin{equation*}
S(t)=\operatorname{Exp}\{-\lambda t\} \ldots \text { where } \ldots \lambda=\text { the hazard rate } \ldots \text { and } \ldots 0 \leq \lambda \leq 1 \tag{1}
\end{equation*}
$$

We want to map the survival rates in Table 1 above to the Exponential distribution. If we define $N$ to be sample size then the equation for the parameter $\lambda$ via a Maximum Log Likelihood MLL) estimation methodology is... [2]

$$
\begin{equation*}
\lambda=\left[\frac{1}{N} \sum_{i=1}^{N} t_{i}\right]^{-1} \tag{2}
\end{equation*}
$$

Using the data in Table 1 above we will define the following parameters...

$$
\begin{equation*}
N=21 \quad \ldots \text { and } \ldots \sum_{i=1}^{N} t_{i}=210 \tag{3}
\end{equation*}
$$

Using Equations (2) and (3) above the MLL estimate of the hazard rate parameter $\lambda$ is...

$$
\begin{equation*}
\lambda=\left[\frac{1}{21} \times 210\right]^{-1}=0.1000 \tag{4}
\end{equation*}
$$

Using Equations (1) and (4) above and the data in Table 1 above the graph of actual survival rates versus estimated survival rates via an Exponential distribution is...

Table 3: Survival Rates

| Birth <br> Year | Actual <br> Rate | Estimate <br> Rate | Est-Act <br> Residual |
| :---: | :---: | :---: | ---: |
| 0 | 1.0000 | 1.0000 | - |
| 1 | 0.8010 | 0.9048 | 0.1038 |
| 2 | 0.6870 | 0.8187 | 0.1317 |
| 3 | 0.6160 | 0.7408 | 0.1248 |
| 4 | 0.5600 | 0.6703 | 0.1103 |
| 5 | 0.4880 | 0.6065 | 0.1185 |
| 6 | 0.4470 | 0.5488 | 0.1018 |
| 7 | 0.4160 | 0.4966 | 0.0806 |
| 8 | 0.3800 | 0.4493 | 0.0693 |
| 9 | 0.3550 | 0.4066 | 0.0516 |
| 10 | 0.3380 | 0.3679 | 0.0299 |
| 11 | 0.3160 | 0.3329 | 0.0169 |
| 12 | 0.2990 | 0.3012 | 0.0022 |
| 13 | 0.2840 | 0.2725 | -0.0115 |
| 14 | 0.2660 | 0.2466 | -0.0194 |
| 15 | 0.2510 | 0.2231 | -0.0279 |
| 16 | 0.2400 | 0.2019 | $-0,0381$ |
| 17 | 0.2300 | 0.1827 | -0.0473 |
| 18 | 0.2220 | 0.1653 | -0.0567 |
| 19 | 0.2130 | 0.1496 | -0.0634 |
| 20 | 0.2040 | 0.1353 | -0.0687 |

Table 4: Survival Rates Mapped To An Exponential Distribution


Note that in the graph above survival rates mapped to an Exponential distribution is not that accurate as survival rates are over-estimated over the first ten years and under-estimated over the last then years.

## Residual Estimation via Linear Regression

Using the data in Table 3 above the graph of the residuals is...


Note that the residuals in the graph above is approximately linear and therefore we will use linear regression to estimate the equation for that line. The Least Squares Estimates (LSE) via the Excel regression tool (the independent variable is time and the dependent variable is the residual value) are...

| Symbol | Description | Value |
| :---: | :--- | ---: |
| R2 | R-squared | 0.96511 |
| $\alpha$ | Constant | 0.15297 |
| $\beta$ | Beta coefficient | -0.01167 |

Using Equations (1) and (4) above and the data in the regression parameters table above the composite survival rate equation is...

$$
\begin{equation*}
S(t)=\operatorname{Exp}\{-\lambda t\}-(\alpha+\beta t) \ldots \text { when } . . t>0 \tag{5}
\end{equation*}
$$

Using Equation (5) above and the data in the regression parameters table above the composite equation for the expected survival rate at time $t$ for our hypothetical problem is...

$$
\begin{align*}
& S(0)=1.0000 \ldots \text { when } \ldots t=0 \\
& S(t)=\operatorname{Exp}\{-0.1000 \times t\}-0.15297+0.01167 \times t \ldots \text { when } \ldots t>0 \tag{6}
\end{align*}
$$

Table 5: Survival Rates


Note that in the graph above the composite survival rate equation (Equation (5) above) gives us a much better (almost perfect) estimate of actual JP Morgan survival rates.

## The Answer To Our Hypothetical Problem

Question: Given that ABC Company has survived to year 3.25 , what is the probability that ABC company will survive until year 7.75?

Using Equation (6) above the unconditional probability that ABC Company survives to year 3.25 is...

$$
\begin{equation*}
\text { Prob }[\text { Survive until year } 3.25]=\operatorname{Exp}\{-0.1000 \times 3.25\}-0.15297+0.01167 \times 3.25=0.6075 \tag{7}
\end{equation*}
$$

Using Equation (6) above the unconditional probability that ABC Company survives to year 7.75 is...

$$
\begin{equation*}
\text { Prob }[\text { Survive until year } 7.75]=\operatorname{Exp}\{-0.1000 \times 7.75\}-0.15297+0.01167 \times 7.75=0.3982 \tag{8}
\end{equation*}
$$

Using Equations (7) and (8) above the answer to the question is... [3]

$$
\begin{equation*}
\text { Prob }[\text { Survive until year } 7.75 \mid \text { Survived to year } 3.25]=\frac{0.3982}{0.6075}=0.6555 \tag{9}
\end{equation*}
$$

## References

[1] Gary Schurman, The Exponential Distribution - The Mathematics, March, 2012.
[2] Gary Schurman, Maximum Likelihood Estimation - The Exponential Distribution, March, 2023.
[3] Gary Schurman, Startup Survival Rates - Discrete-Time Survival Rate Curve, March, 2023.

