

Practical ML Tutorial: Part I

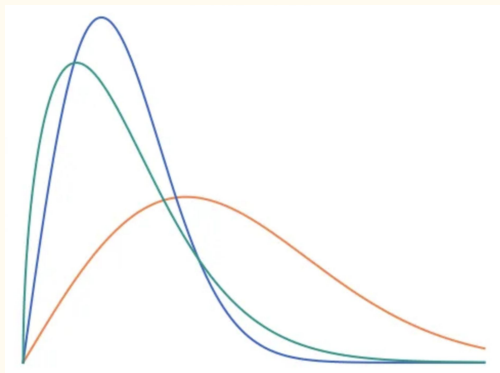
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SEEMAPLD2023, George Williams

Agenda

Part I

- AI Trends
- ML Basics
- Survival Analysis
- Hands-On Programming



Part II

- AI Hardware
- PyTorch Basics
- Computer Vision
- Hands-On Programming



AI Trends

- ChatGPT
- Stable Diffusion
- Deep Fakes
- Alpha Fold
- Foundation Models





**STABLE
DIFFUSION**



DALLE 2



MIDJOURNEY

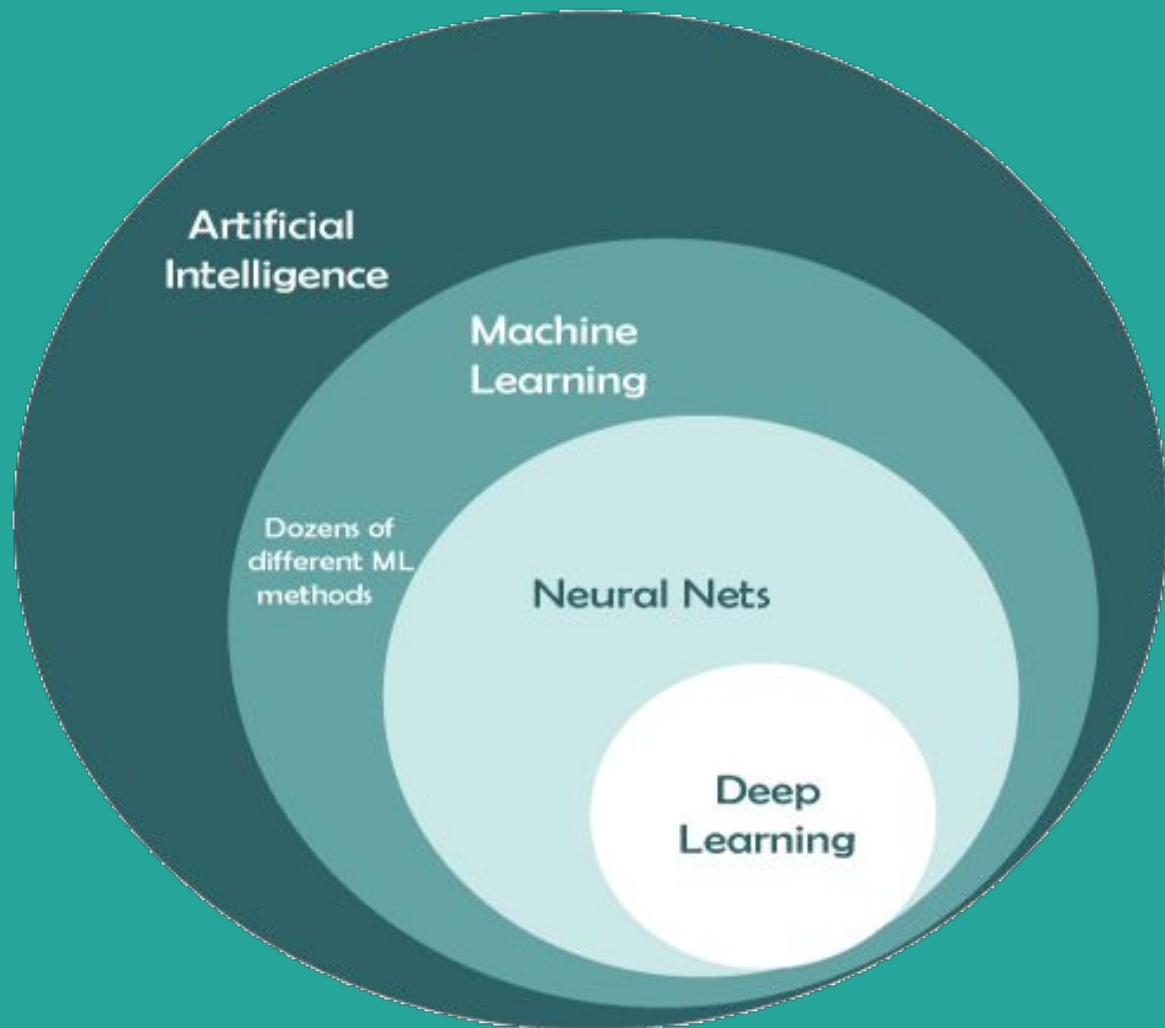
nature



DeepMind

Machine Learning Basics





Machine Learning Basics

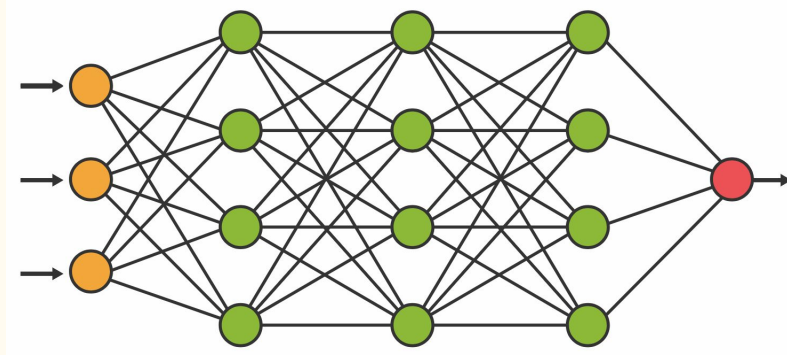


Data

Machine Learning Basics



Data

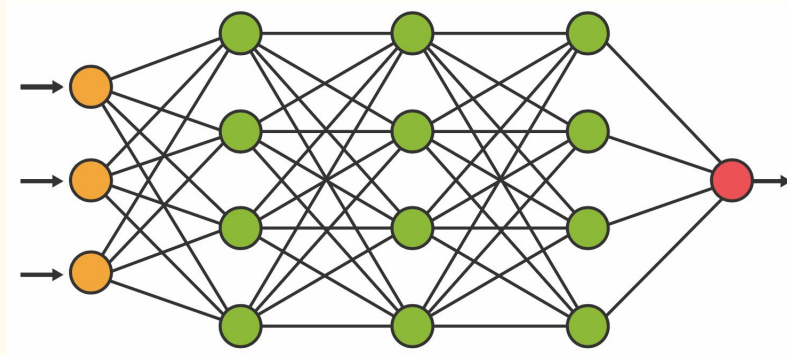


Model

Machine Learning Basics



Data

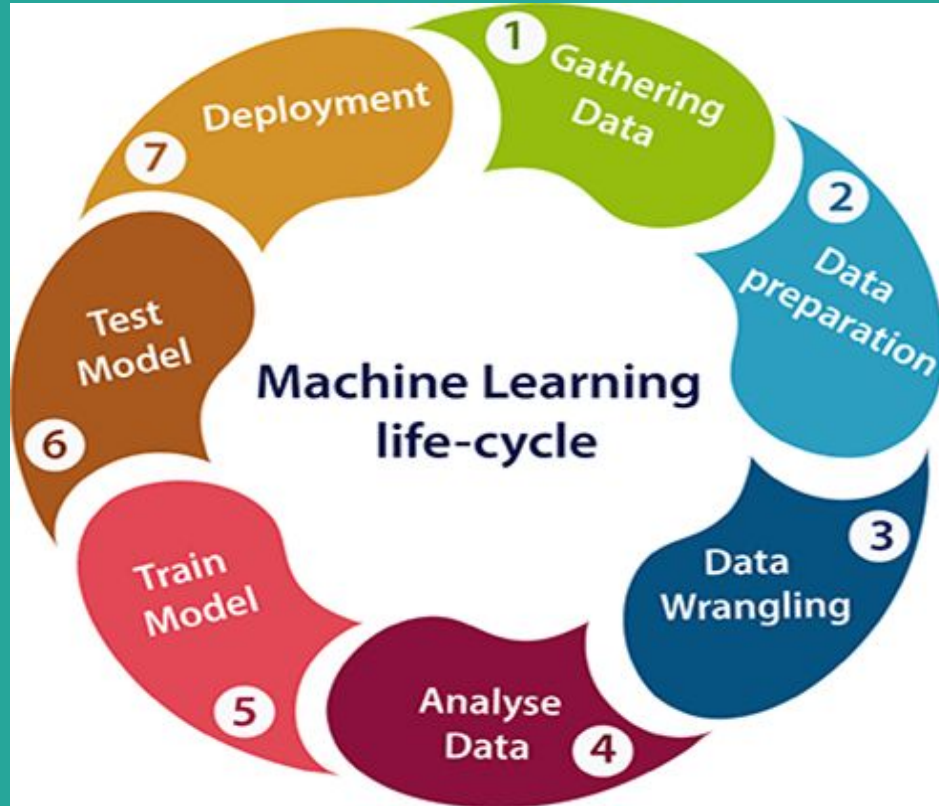


Model



Eval

Machine Learning “In The Wild”



Survival Analysis

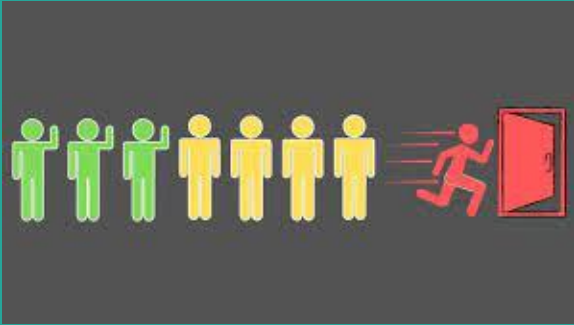
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Tries To Answer The
Question:

When Will It End?

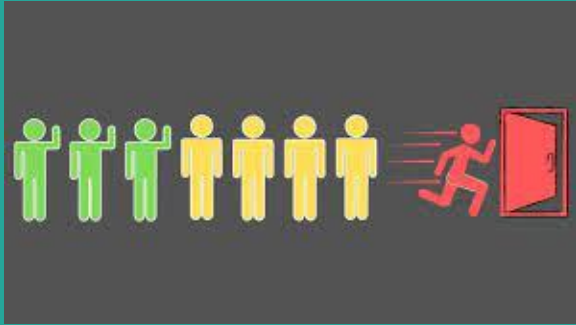
*When Will **It** End?*

*When Will **It** End?*



Customer Churn

When Will *It* End?

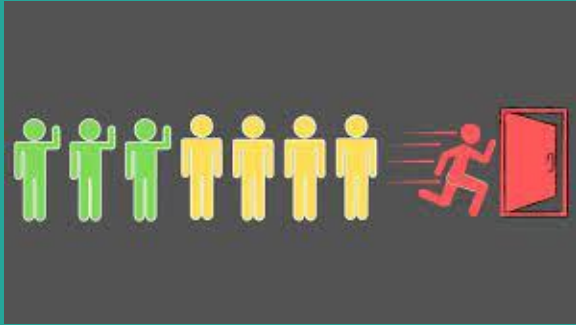


Customer Churn



Health Outcomes

When Will *It* End?



Customer Churn

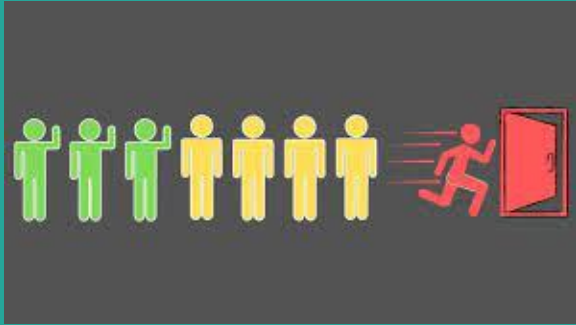


Health Outcomes



Machine Failure

When Will *It* End?



Customer Churn



Health Outcomes



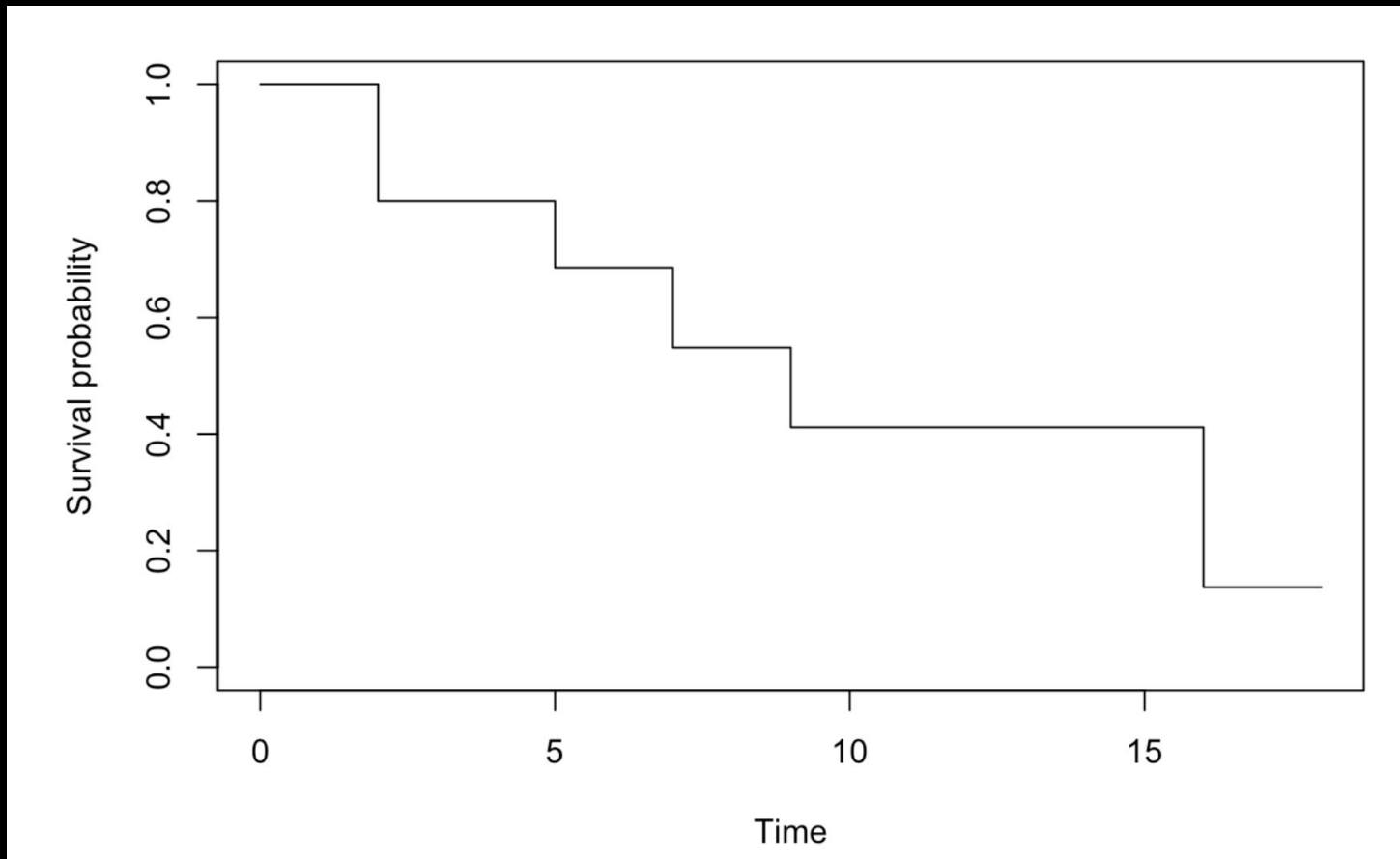
Machine Failure

“Mission Critical” Predictions In Multi-Billion Dollar Industries!

Censoring
period^{free}
time
event

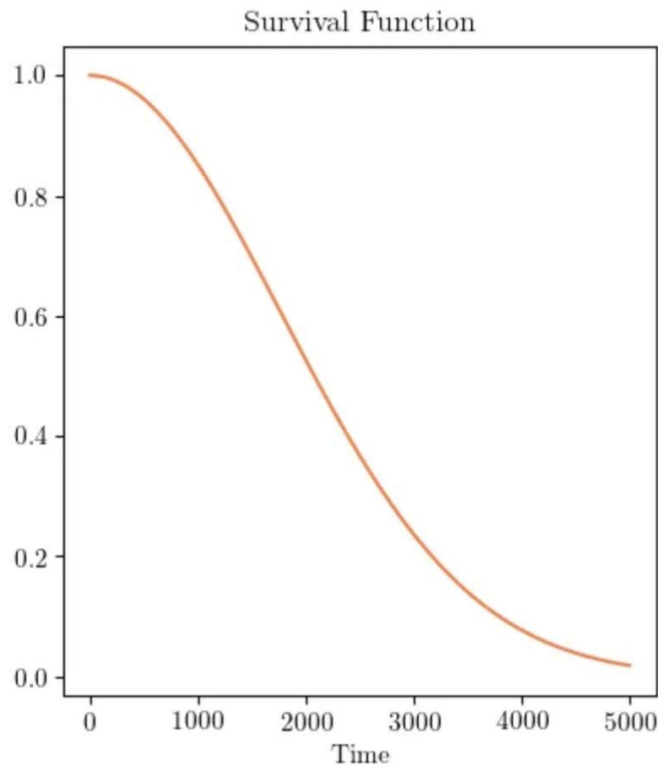
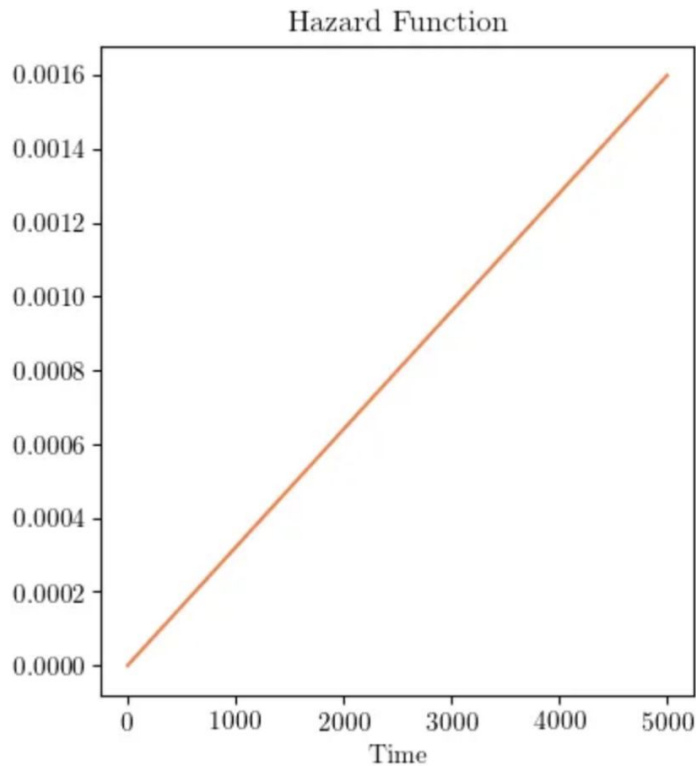
Survival^{drops}
follow-up
observed
follow
probability^{interest}
loss^{right}
censoring
observation

Survival Prediction Is Probabilistic...



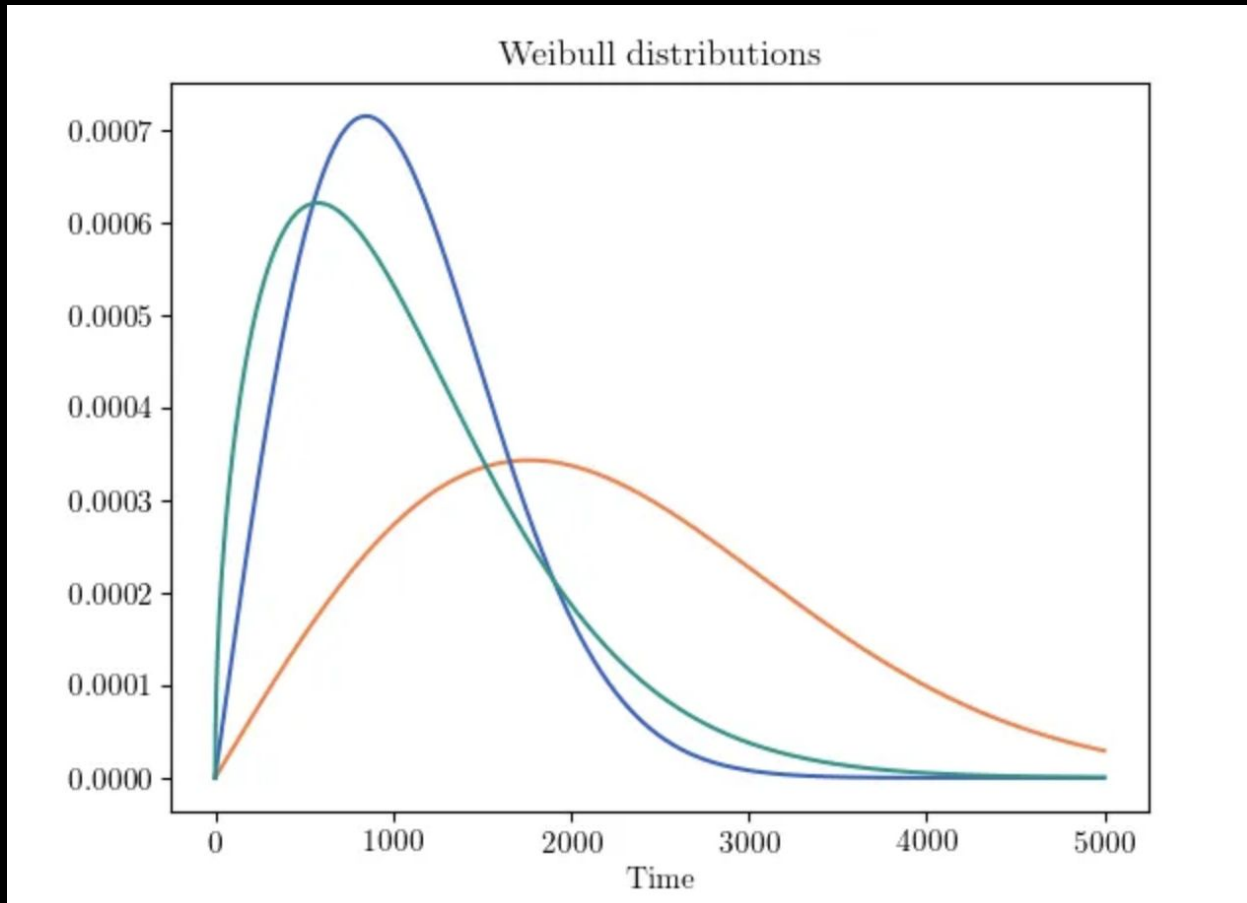
Survival Prediction Is Probabilistic...

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



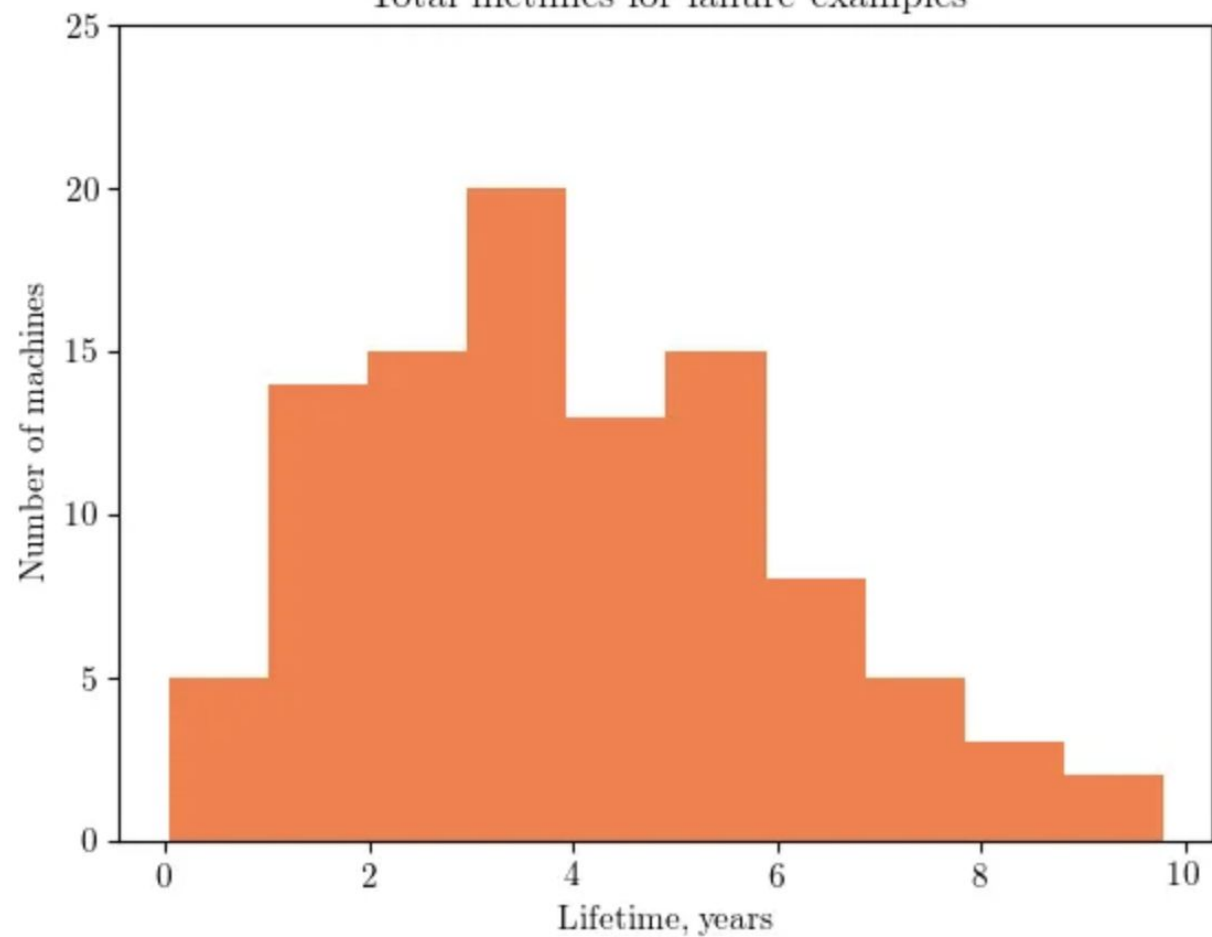
...and we can use the data we have to predict the remaining life of each of a group of machines. To understand how this can be done, it's easiest

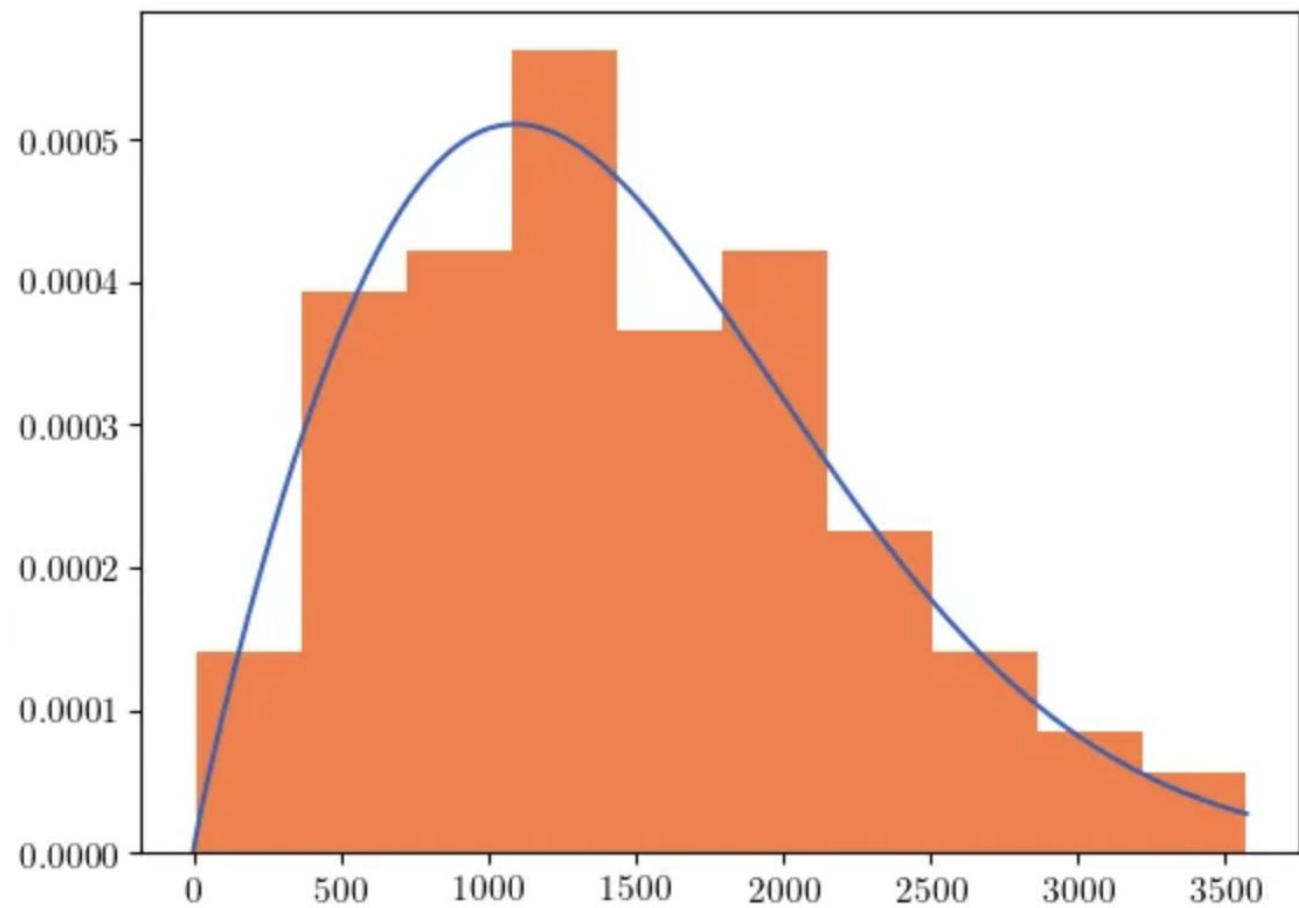
Survival Prediction Is Probabilistic...



Shape
Scale

Total lifetimes for failure examples





Machine Learning Approach

Beyond Curve-Fitting

- Instead of just fitting curves, can we learn from data?
 - There may be multiple underlying distributions
 - Data may be “censored”
 - Best of both worlds: distribution-driven + learning-driven
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Let's Start Coding!

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