# Mortality and Machine Learning: A Glance at Death, Holistically and Precisely

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### Introduction

Research on mortality typically explores factors from either a biological or non-biological perspective. Academics have identified an abundance of theoretical risk factors which 'contribute' to and are statistically associated with death. However, we recognise several areas for improvement in this body of work:

**Predictive**: There is little evidence concerning the predictability of death. The use of predictive frameworks is not a common research design in social sciences, presenting an area for further exploration.

**Holistic**: Existing studies generally draw evidence from one or two disciplines. A holistic view of risk factors, particularly from a non-biological perspective, is rarely explored.

**Precise**: As methodologies advance, we present several concerns regarding prediction precision in health-related outcomes.

Evaluation of Model Performance in Death Prediction over 10 seeds									
		HRS			SHARE			ELSA	
Metrics	SL	LightGBM	LR	SL	LightGBM	LR	$_{\rm SL}$	LightGBM	LR
IMV	0.200	0.204	0.211	0.069	0.072	-0.024	0.016	0.015	0.01
ROC AUC	0.822	0.825	0.831	0.795	0.800	0.448	0.908	0.909	0.89
PR AUC	0.689	0.687	0.707	0.497	0.507	0.161	0.262	0.246	0.25
EFRON $R^2$	0.288	0.290	0.307	0.196	0.204	-0.050	0.096	0.080	0.13
FFC $R^2$	0.562	0.564	0.574	0.446	0.452	0.277	0.224	0.210	0.25
IP	0.293	0.293	0.293	0.196	0.196	0.196	0.057	0.057	0.05
HRS + SHARE			Е	HRS + ELSA			SHARE + ELSA		
Metrics	SL	LightGBM	LR	SL	LightGBM	LR	SL	LightGBM	LR
IMV	0.106	0.108	0.108	0.120	0.122	0.116	0.065	0.067	0.06
ROC AUC	0.808	0.810	0.810	0.860	0.862	0.852	0.833	0.837	0.82
PR AUC	0.580	0.585	0.585	0.624	0.625	0.629	0.481	0.497	0.48
EFRON $R^2$	0.238	0.241	0.242	0.304	0.307	0.306	0.217	0.228	0.21
FFC $R^2$	0.511	0.513	0.513	0.528	0.530	0.530	0.410	0.419	0.41
IP	0.236	0.236	0.236	0.211	0.211	0.211	0.157	0.157	0.15

Density 20

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**Precise: Seed Variability** 



Methods and Data						
Datasets	Predictive Models	<ul> <li>Holistic Comparisons</li> <li>Quantifying Precision</li> </ul>				
1.HRS	a.Super Learner	SHAP Decomposition     1,2,4				
2.SHARE	Train b.LightGBM	Domain Contribution				

- 1. U.S. Health and Retirement Study Death Prevalence: 30.0%
- 2. Survey of Health, Ageing and Retirement in Europe Death Prevalence: 19.5%
- 3. English Longitudinal Study of Ageing Death Prevalence: 5.40%

#### Domain of Risk Factors:

Demography, Socioeconomics, Psychology, Adulthood Adversity, Childhood Adversity, Social Connections, Health Behaviours





1162 21141 18672 28196 11721

**Precise: Asymptotic** 

162 21147 24672 18196 3172

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Efron R2

PR-AUC Scor





#### Outcome Distribution Characteristics of 10000 seeds

Туре	Metric/Predictor	Mean	Min	Max	Standard Deviation
Evaluation Metrics	PR-AUC	0.625	0.597	0.651	0.007
	IMV	0.119	0.109	0.131	0.003
	Efron $R^2$	0.288	0.266	0.309	0.006
	FFC $R^2$	0.531	0.501	0.558	0.007
SHAP	Current Smoker	0.141	0.090	0.187	0.012
	Age	0.965	0.926	1.004	0.011
	Low/No Moderate Activity	0.115	0.071	0.15	0.010
	Male	0.229	0.196	0.272	0.010

#### Note: FFC $R^2 = 1 - \frac{\sum (y_i - \pi_i)^2}{\sum (y_i - \overline{y_{i-r_i}})^2}$

 $IMV = \frac{\omega_1 - \omega_0}{\omega_0}$ ,  $\omega_0$  is in-sample prevalence,  $\omega_1$  is the probability that defines the entropy of the model

## **Holistic: Domain Contribution**



Visit booth 312 and scan the code to find out more

