

# Heart Failure

Survival Analysis

April 5, 2023

**SURVIVAL ANALYSIS MATH 4130K**

Department of Mathematics

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What we'll discuss  
this afternoon

# Presentation Overview

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Research Background

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Data Overview

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Methodology

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Analysis Walkthrough

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Interpretation & Limitations

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Conclusion

# What is heart failure and causes it?

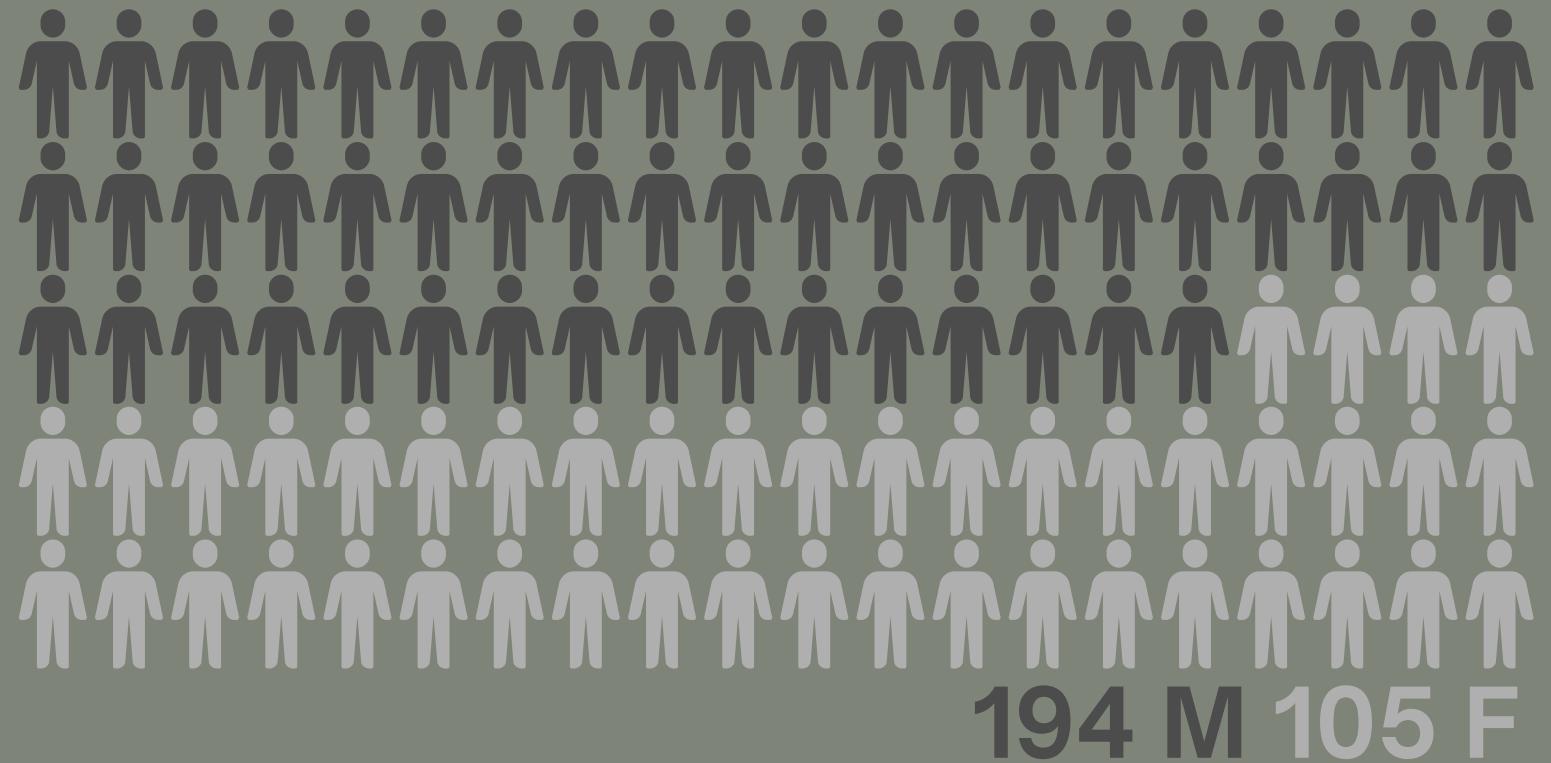
A medical condition in which the heart is unable to pump enough blood to meet the body's demands

The main reasons for heart failure include coronary heart disease, diabetes, high blood pressure, and other diseases like HIV

This can occur when the heart muscle is weakened or stiffened, which makes it harder for the heart to fill with blood or pump blood out to the body.

# DATA OVERVIEW

[Source: UCI LEARNING REPOSITORY]



DATA COLLECTION :

INSTITUTE OF CARDIOLOGY AND ALLIED  
HOSPITAL IN FAISALABAD

COLLECTION TIME :

APRIL-DECEMBER 2015

**299 Subjects**

203 Censored 96 Observed

**13 Clinical  
Features**

# 13 Clinical Features

AGE

ANAEMIA

PLATELETS

SEX

CREATININE PHOSPHOKINASE

SERUM CREATININE

DIABETES

EJECTION FRACTION

SERUM SODIUM

SMOKING

HIGH BLOOD PRESSURE

TIME

DEATH [TARGET EVENT]

# 13 Clinical Features

**AGE**

(numerical)

**SEX**

(binary)

**DIABETES**

(boolean)

**SMOKING**

(boolean)

**TIME**

follow-up period (days)

**ANAEMIA**

decrease of red blood cells or hemoglobin (boolean)

**CREATININE PHOSPHOKINASE**

CPK enzyme level in blood msg/L

**EJECTION FRACTION**

percentage of blood leaving the heart at each contraction

**HIGH BLOOD PRESSURE**

(boolean)

**DEATH [TARGET EVENT]**

if the patient deceased during the follow-up period (boolean)

**PLATELETS**

platelets in the blood (kiloplatelets/mL)

**SERUM CREATININE**

level of serum creatinine in the blood (mg/dL)

**SERUM SODIUM**

level of serum sodium in the blood (mEq/L)

# 13 Clinical Features

**AGE**  
(numerical)

**SEX**  
(binary)

**DIABETES**  
(boolean)

**SMOKING**  
(boolean)

**TIME**  
follow-up period (days)

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decrease of red blood cells or  
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CPK enzyme level in blood msg/L

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percentage of blood leaving the  
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**HIGH BLOOD PRESSURE**  
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level of serum sodium in the blood  
(mEq/L)

# Features of Interest

[TARGET EVENT]

DEATH

AGE            SERUM CREATININE (SCR)

SEX            EJECTION FRACTION (EF)

TIME            HIGH BLOOD PRESSURE (HBP)

# Methodology

## Kaplan-Meier PL Estimators

$$\hat{S}(t) = \prod_{j:t_j \leq t} (1 - d_j/n_j)$$
 where  $d_j/n_j$  is the estimated hazard for  $j$ 

01

Non-parametric Method

02

Utilized in comparing the survival rates between different levels within categorical covariates

03

Variance of the KM curve approximated by Greenwood's formula with confidence interval at each value of  $t$

# Cox Proportional Hazard Model

$$\lambda_i(t, x_i) = \lambda_0(t) \exp\{x_i^T \beta\}$$

where  $\lambda_0(t)$  is the baseline hazard

## Proportional Hazard Assumption

Hazard ratio between two individuals with different predictors does not depend on time.

$$\frac{\lambda_2(t, x_2)}{\lambda_1(t, x_1)} = \frac{\exp\{x_2^T \beta\}}{\exp\{x_1^T \beta\}}$$

# Model Assumptions

## Methods of Verification

### 01 Log-Log Plots

Using the log-log plot on the survival functions of two individuals to check for parallel trends

The difference between the individuals are independent of time

### 02 Schoenfeld Residuals

Using the `zph()` and `cox.zph()` functions to check for proportionality assumption in R

Having a small p-values would indicate undesired time dependent coefficients

# Analysis

## First Steps

- 55 to 70% – Normal heart function.
- 40 to 55% – Below normal heart function.  
Can indicate previous heart damage from heart attack or cardiomyopathy.
- >75% – Can indicate a heart condition like hypertrophic cardiomyopathy, a common cause of sudden cardiac arrest.
- <=40% – May confirm the diagnosis of heart failure.

Prepare and clean data through:

### 01 Data Wrangling

### 02 Variable Manipulation

*Ejection Fraction (EF)* is converted into a categorical variable with 4 factor levels, according to Penn Medicine.

### 03 Standardize Continuous Variables



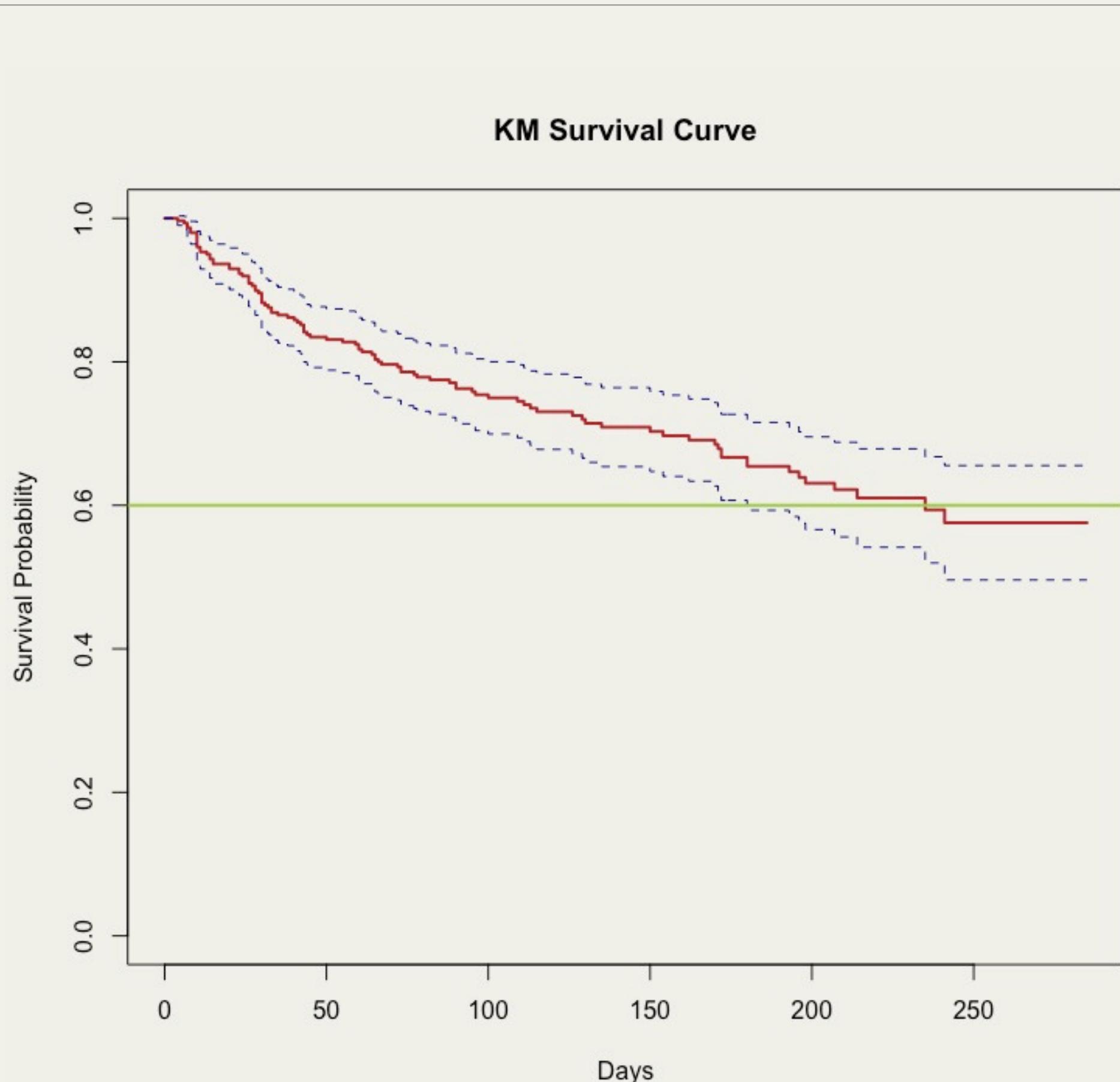
Penn Medicine

# Analysis

## Kaplan-Meier Survival Curve

### The survival curve before modelling

Shows a gradual trend with a relatively high survival probability at just below 60% over long survival times



# Cox Proportional Hazard Model

## Full Model

$$\hat{\lambda}(t, x_i) = \hat{\lambda}_0(t) \exp\{x_{i1}\hat{\beta}_{age} + x_{i2}\hat{\beta}_{CPK} + x_{i3}\hat{\beta}_{EF1} + x_{i4}\hat{\beta}_{EF2} + x_{i5}\hat{\beta}_{EF3} + x_{i6}\hat{\beta}_{platelets} + x_{i7}\hat{\beta}_{SCR} + x_{i8}\hat{\beta}_{SSO} + x_{i9}\hat{\beta}_{anaemia} + x_{i10}\hat{\beta}_{diabetes} + x_{i11}\hat{\beta}_{HBP} + x_{i12}\hat{\beta}_{sex} + x_{i13}\hat{\beta}_{smoking}\}$$

## Regression Output from R

	term	estimate	std.error	z statistic	p.value
1	age	0.5096	0.1086	4.6930	0.0000
2	CPK	0.1897	0.0992	1.9128	0.0558
3	EF1	-0.2167	0.3194	-0.6784	0.4975
4	EF2	-0.8666	0.4369	-1.9837	0.0473
5	EF3	-12.5179	2168.8692	-0.0058	0.9954
6	platelets	-0.0360	0.1101	-0.3271	0.7436
7	SCR	0.3090	0.0689	4.4838	0.0000
8	SSO	-0.2560	0.0963	-2.6570	0.0079
9	anaemia	0.4226	0.2170	1.9471	0.0515
10	diabetes	0.1271	0.2231	0.5696	0.5689
11	HBP	0.4978	0.2172	2.2923	0.0219
12	sex	-0.1207	0.2529	-0.4773	0.6331
13	smoking	0.1641	0.2536	0.6472	0.5175

# Reduced Model

Four significant covariates are selected for further analysis:

**Age**

**EF** (ejection fraction)

**SCR** (serum creatinine)

**HBP** (high blood pressure)

EF and HBP are categorical indicator variables\*

The **two-way interaction effects** between them must be considered

$$\hat{\lambda}_0(t) \exp\{x_{i1}\hat{\beta}_{age} + x_{i2}\hat{\beta}_{EF1} + x_{i3}\hat{\beta}_{EF2} + x_{i4}\hat{\beta}_{EF3} + x_{i5}\hat{\beta}_{SCR} + x_{i6}\hat{\beta}_{HBP} \\ + x_{i7}\hat{\beta}_{EF1*HBP} + x_{i8}\hat{\beta}_{EF2*HBP} + x_{i9}\hat{\beta}_{EF3*HBP}\}$$

# Reduced Model

## Regression Output

**Age, HBP, and SCR** are important predictors of dying of heart failure

The relationship between **EF** and the death is complex and depends on the presence of **HBP**

term	coef	exp(coef)	std.error	z statistic	p-value
age	0.45	1.57	0.11	4.21	0.00
EF1	-0.28	0.75	0.44	-0.65	0.52
EF2	-0.05	0.95	0.44	-0.12	0.90
EF3	-12.76	0.00	2094.77	-0.01	1.00
HBP	0.64	1.89	0.24	2.72	0.01
SCR	0.40	1.49	0.07	5.72	0.00
EF1:HBP	-0.15	0.86	0.63	-0.24	0.81
EF2:HBP	-2.73	0.07	1.14	-2.39	0.02
EF3:HBP	NA	NA	0.00	NA	NA

EF Frequency Table

	$\leq 40\%$	40 - 55%	55 - 70%	>70%
EF Level	0	1	2	3
Counts	219	44	35	1

**NA** Values in the third interaction term due to the fact that there is only **one observation** in the third level of the variable **EF**

# Interpretations

## Hazard Ratio

EF Frequency Table

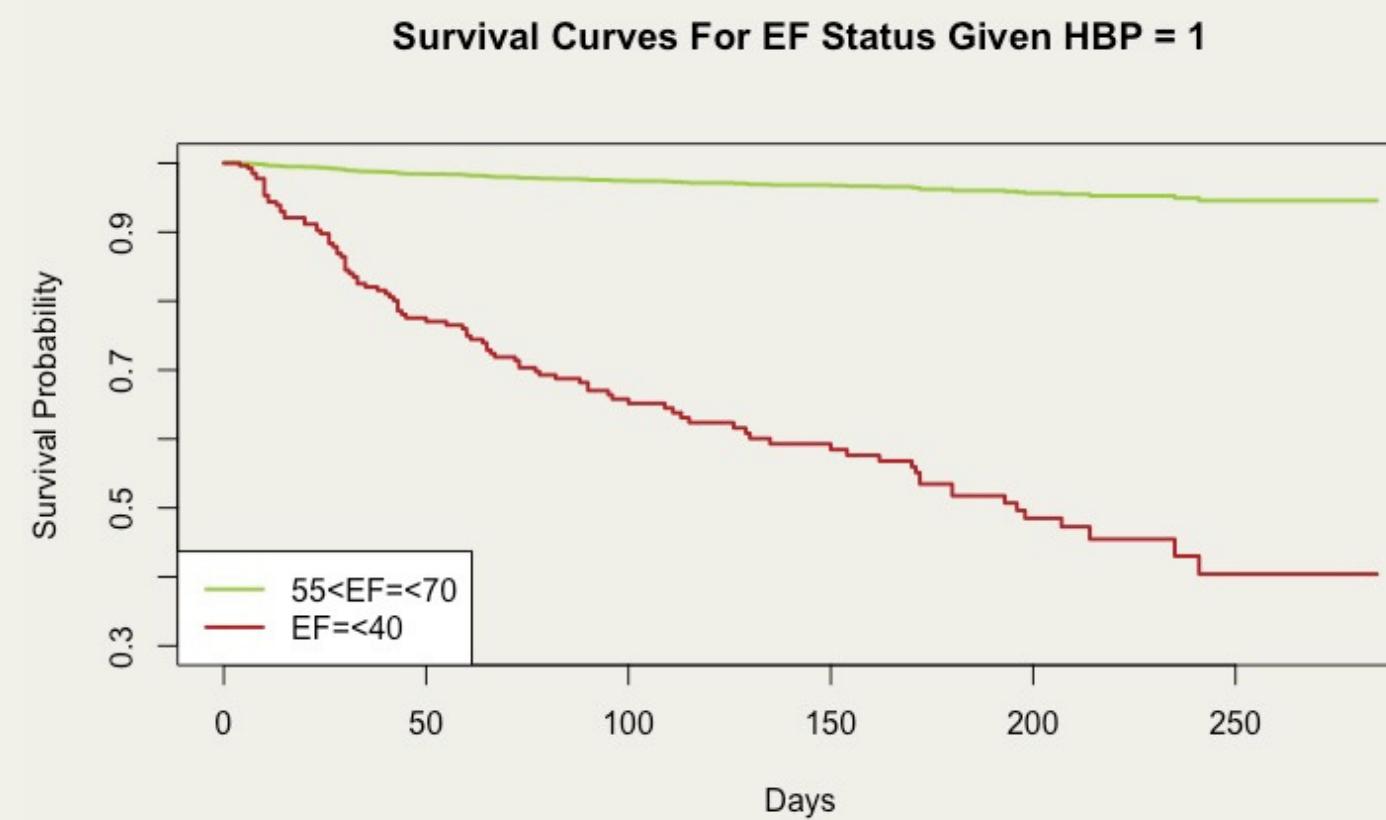
	$\leq 40\%$	40 - 55%	55 - 70%	>70%
EF Level	0	1	2	3
Counts	219	44	35	1

Term	$\exp\{\hat{\beta}_j\}$	HR	Implication
age	1.5659208	$HR > 1$	Hazard increases by a multiple of $\hat{\beta}_j$ per unit increase in age
EF1	0.7541553	$HR < 1$	Hazard decreases by a multiple of $\hat{\beta}_j$ per unit increase in EF1
EF2	0.9476188	$HR < 1$	Hazard decreases by a multiple of $\hat{\beta}_j$ per unit increase in EF2
EF3	0.0000029	$HR < 1$	Hazard decreases by a multiple of $\hat{\beta}_j$ per unit increase in EF3
HBP	1.8939248	$HR > 1$	Hazard increases by a multiple of $\hat{\beta}_j$ per unit increase in HBP
SCR	1.4947745	$HR > 1$	Hazard increases by a multiple of $\hat{\beta}_j$ per unit increase in SCR
EF1:HBP	0.8617743	$HR < 1$	Hazard decreases by a multiple of $\hat{\beta}_j$ per unit increase in EF1:HBP
EF2:HBP	0.0650996	$HR < 1$	Hazard decreases by a multiple of $\hat{\beta}_j$ per unit increase in EF2:HBP

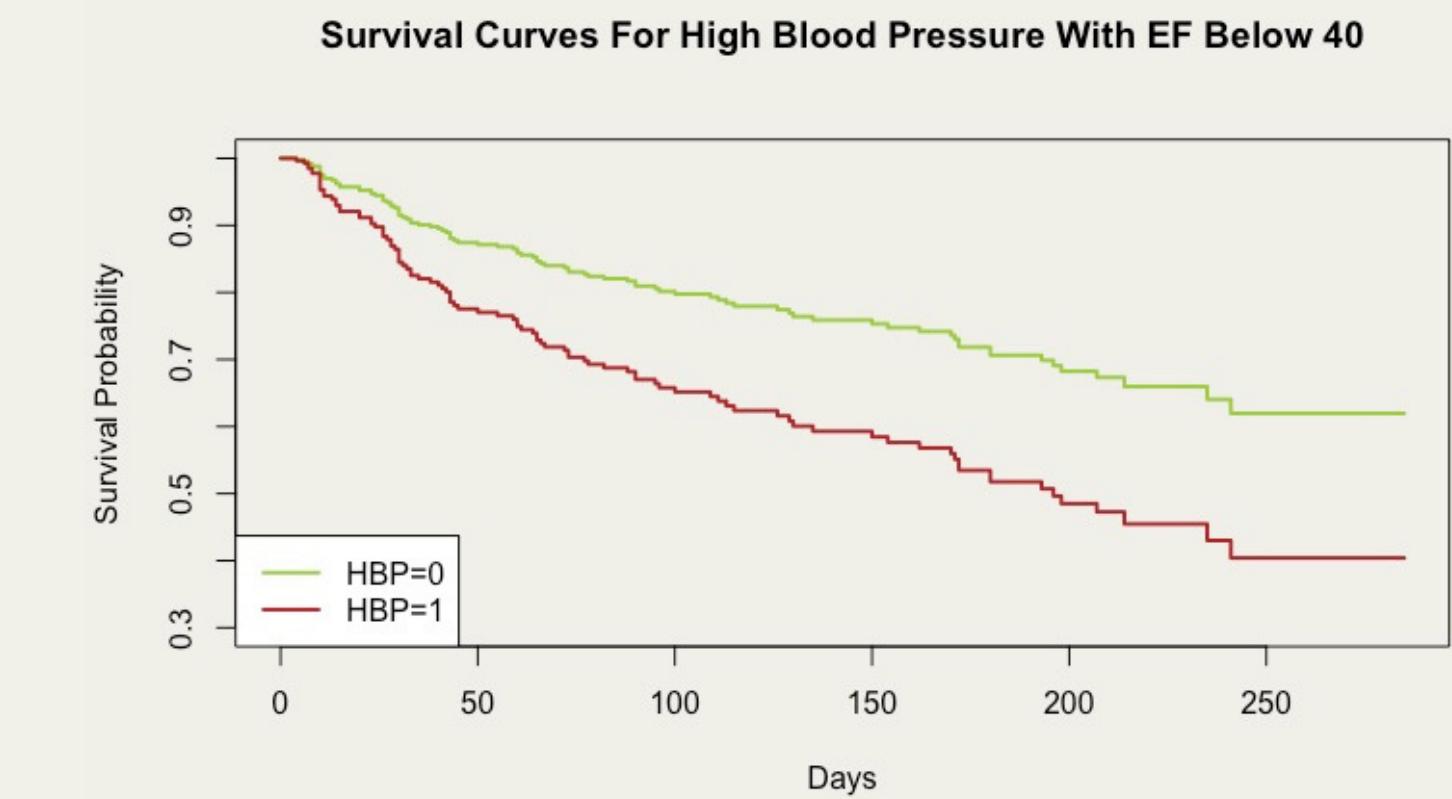
# Conditional Survival Curves

Holding Age and SCR constant...

The presence of HBP



The presence of EF below of 40%

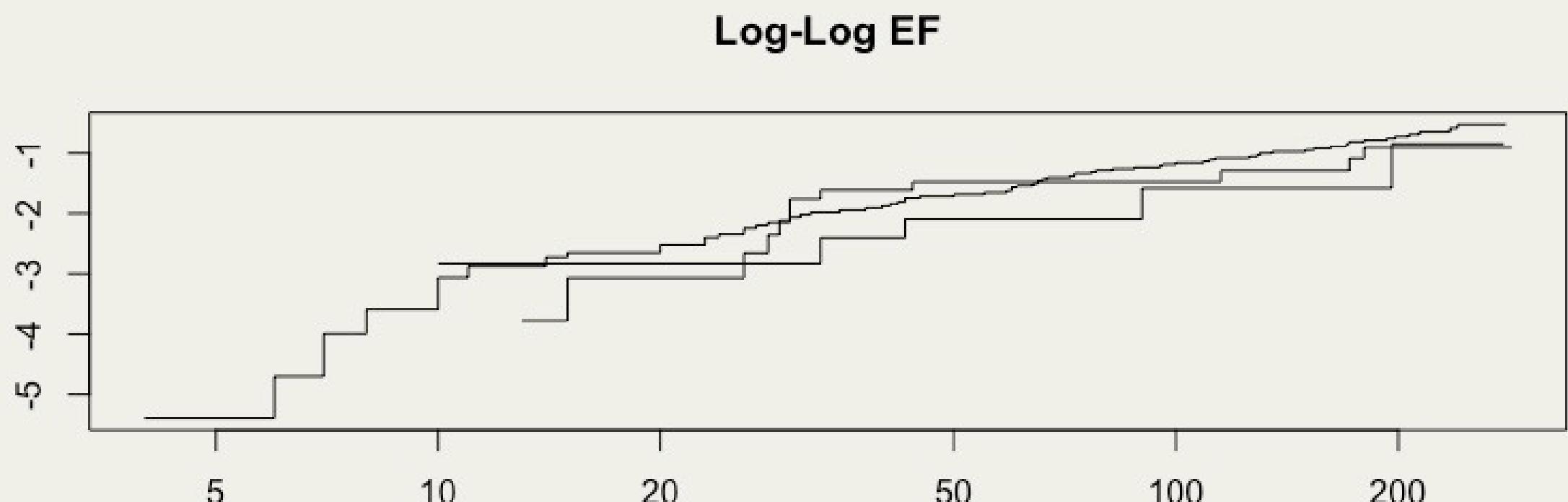
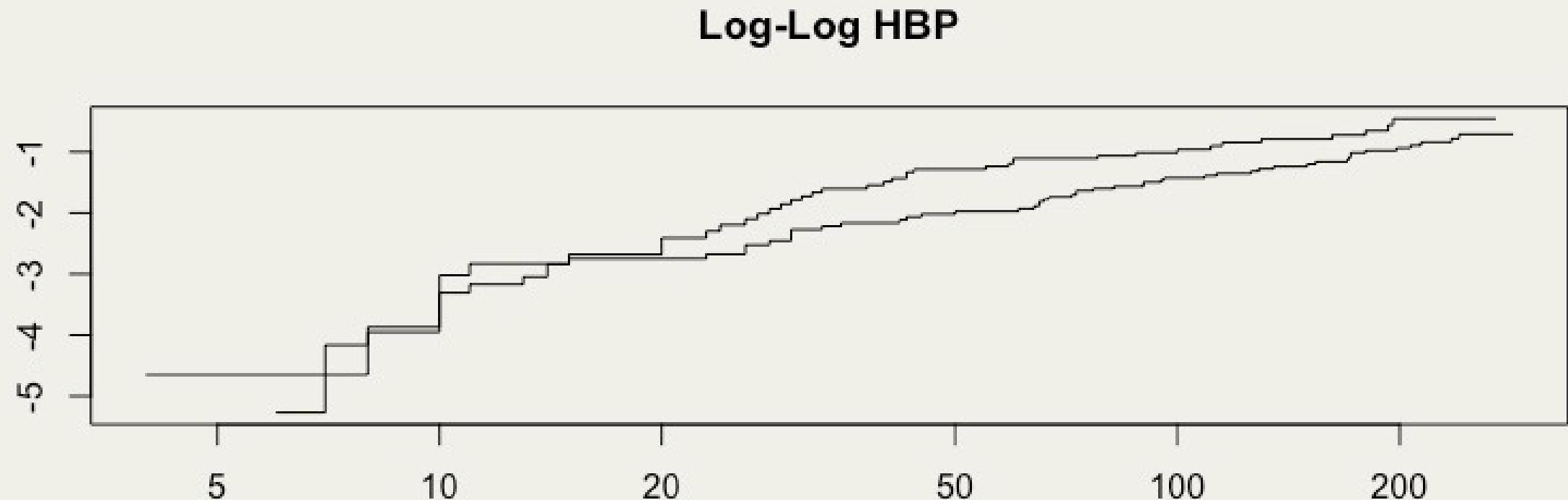


# Verifying Assumptions

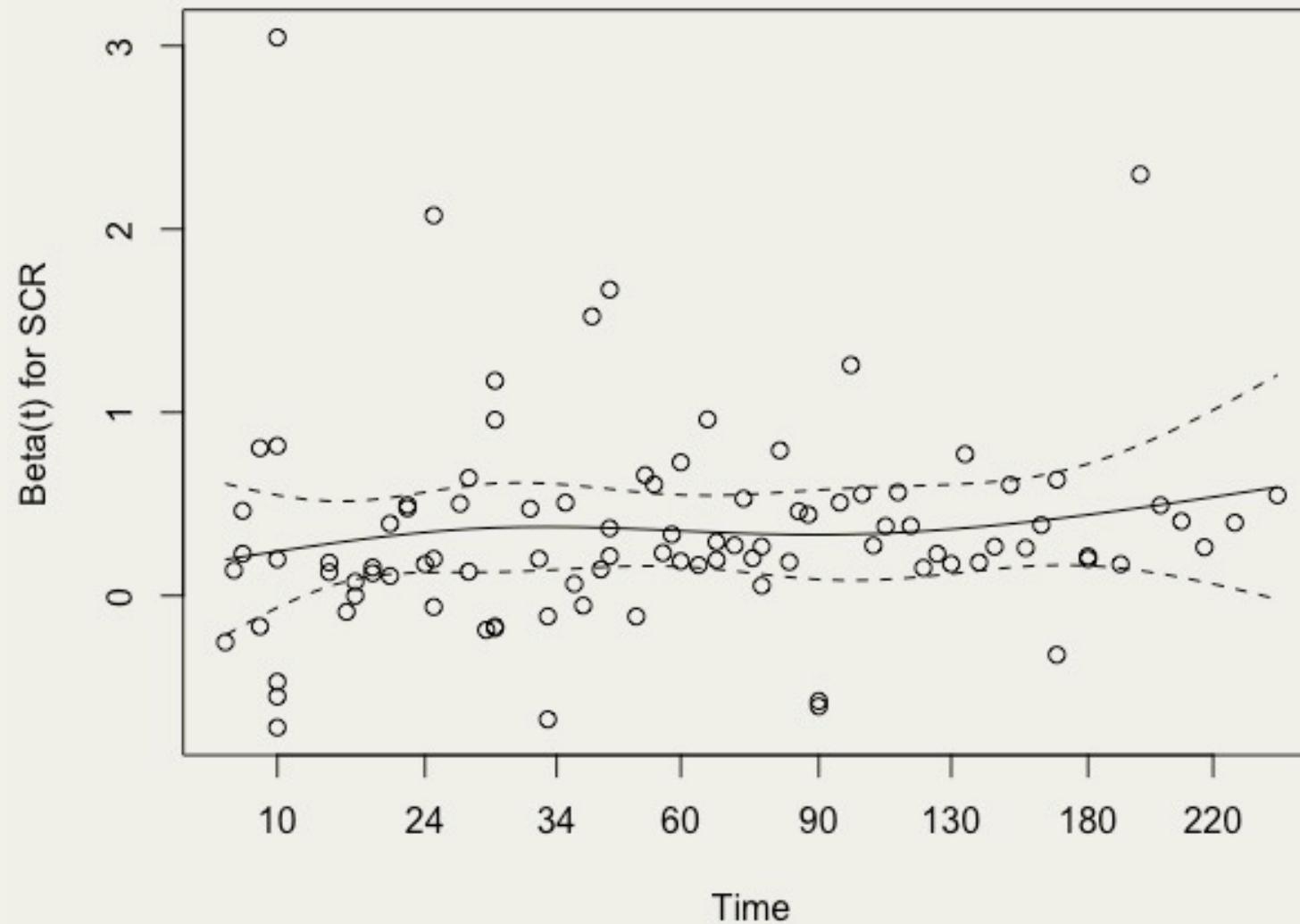
## 01 Log-Log Plots

We proceed by checking the assumptions of the two categorical variables **HBP** and **EF**

	$\leq 40\%$	40 - 55%	55 - 70%	>70%
EF Level	0	1	2	3
Counts	219	44	35	1



# Chi-Squared Table



		$\chi^2$	df	p
	age	0.3067	1	0.58
	EF	0.0637	3	1.00
	HBP	0.2038	1	0.65
	SCR	0.8187	1	0.37
	GLOBAL	1.7389	6	0.94

	$\chi^2$	df	p
age	0.2175	1.0000	0.6409
EF	0.1685	3.0000	0.9825
HBP	0.1488	1.0000	0.6997
SCR	3.8766	1.0000	0.0490
EF:HBP	3.0418	2.0000	0.2185
GLOBAL	5.9843	8.0000	0.6490

# LIMITATIONS

01

Stratified Model Without  
Interaction & Assumption  
Inconsistencies

02

Experimental Design Limitations  
& Sample Bias

03

Lead Time Bias

# Stratified Model

Assumptions regarding Schoenfeld Residuals are all met

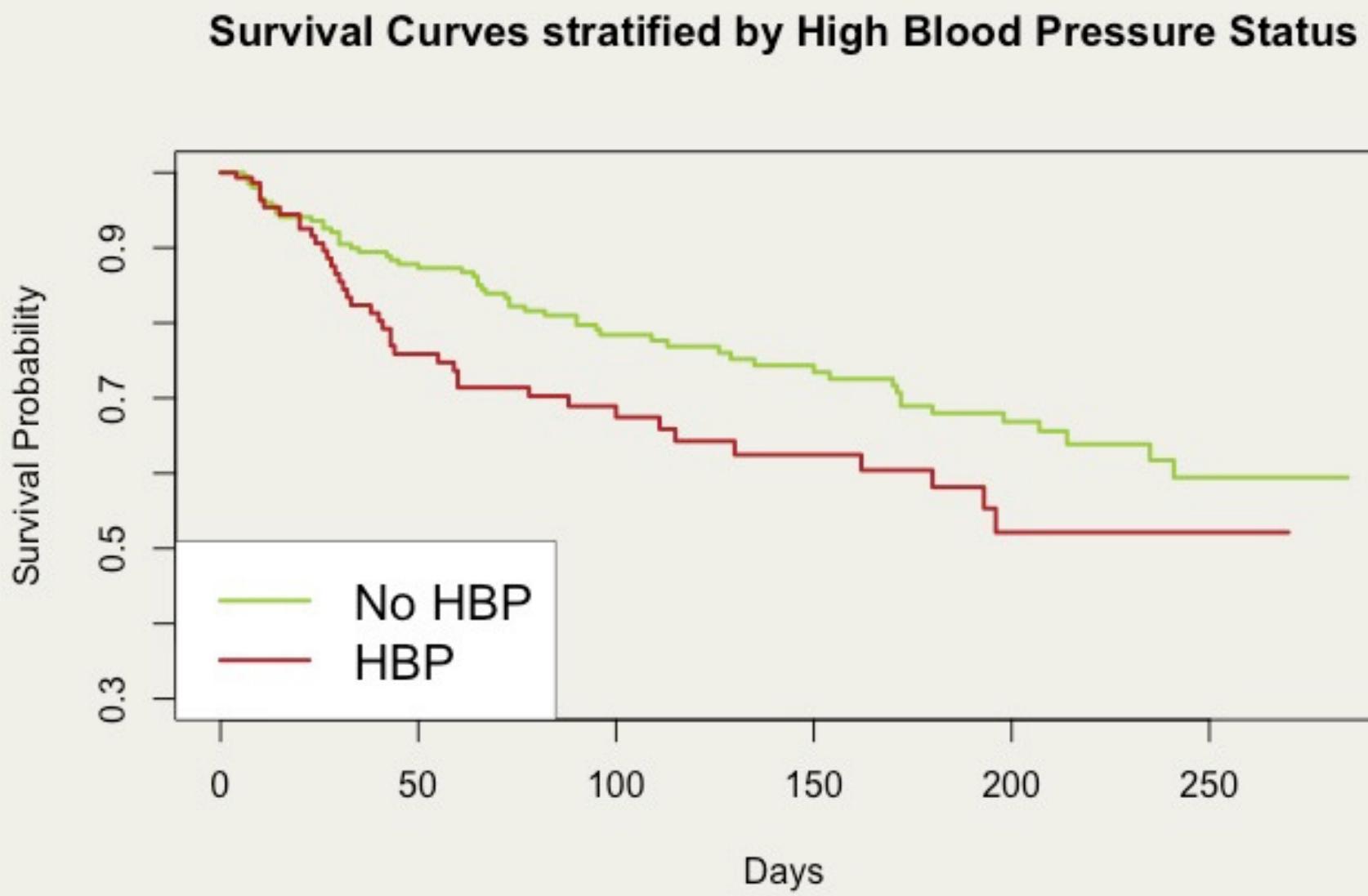
Two-way interaction effects are no longer present

Stratifying by HBP eliminates it from parametric portion of our model

$$\hat{\lambda}_0(t) \exp\{x_{i1}\hat{\beta}_{age} + x_{i2}\hat{\beta}_{EF1} + x_{i3}\hat{\beta}_{EF2} + x_{i4}\hat{\beta}_{EF3} + x_{i5}\hat{\beta}_{SCR}$$

# Stratified Model

$$\hat{\lambda}_0(t) \exp\{x_{i1}\hat{\beta}_{age} + x_{i2}\hat{\beta}_{EF1} + x_{i3}\hat{\beta}_{EF2} + x_{i4}\hat{\beta}_{EF3} + x_{i5}\hat{\beta}_{SCR}\}$$



		$\chi^2$	df	p
	age	0.486	1	0.49
	EF	0.066	3	1.00
	SCR	0.480	1	0.49
	GLOBAL	1.7389	5	0.94

	term	estimate	std.error	statistic	p.value
1	age	0.49	0.11	4.67	0.00
2	EF1	-0.37	0.31	-1.18	0.24
3	EF2	-0.97	0.43	-2.23	0.03
4	EF3	-12.64	2122.89	-0.01	1.00
5	SCR	0.35	0.07	5.17	0.00

# Experimental Design

Small Sample Size  
Observational Study  
No Randomization

## Inclusion Criteria

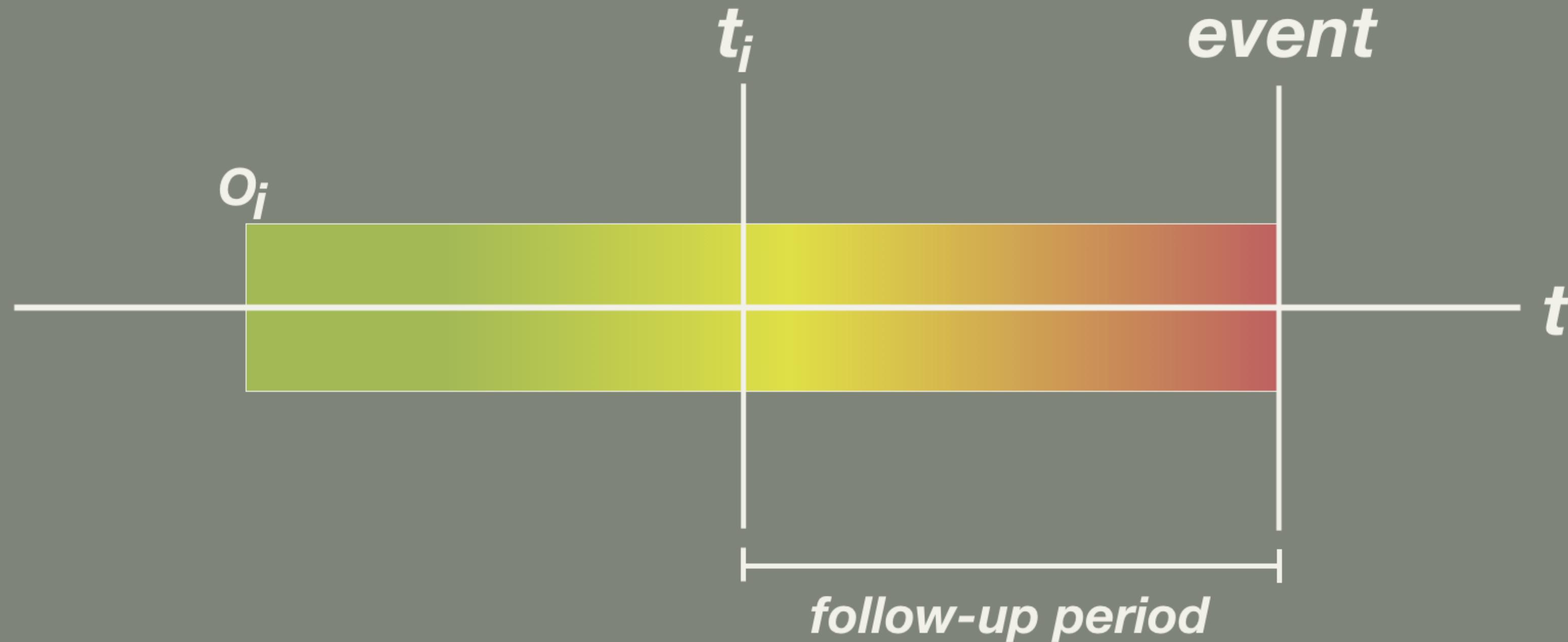
All the patients were aged 40 years or above, having left ventricular systolic dysfunction, belonging to NYHA class III and IV.

## Selection Bias

All the patients were selected in 1 hospital, of 1 geographical region.

# LEAD-TIME BIAS

Chronic vs. Infectious Disease

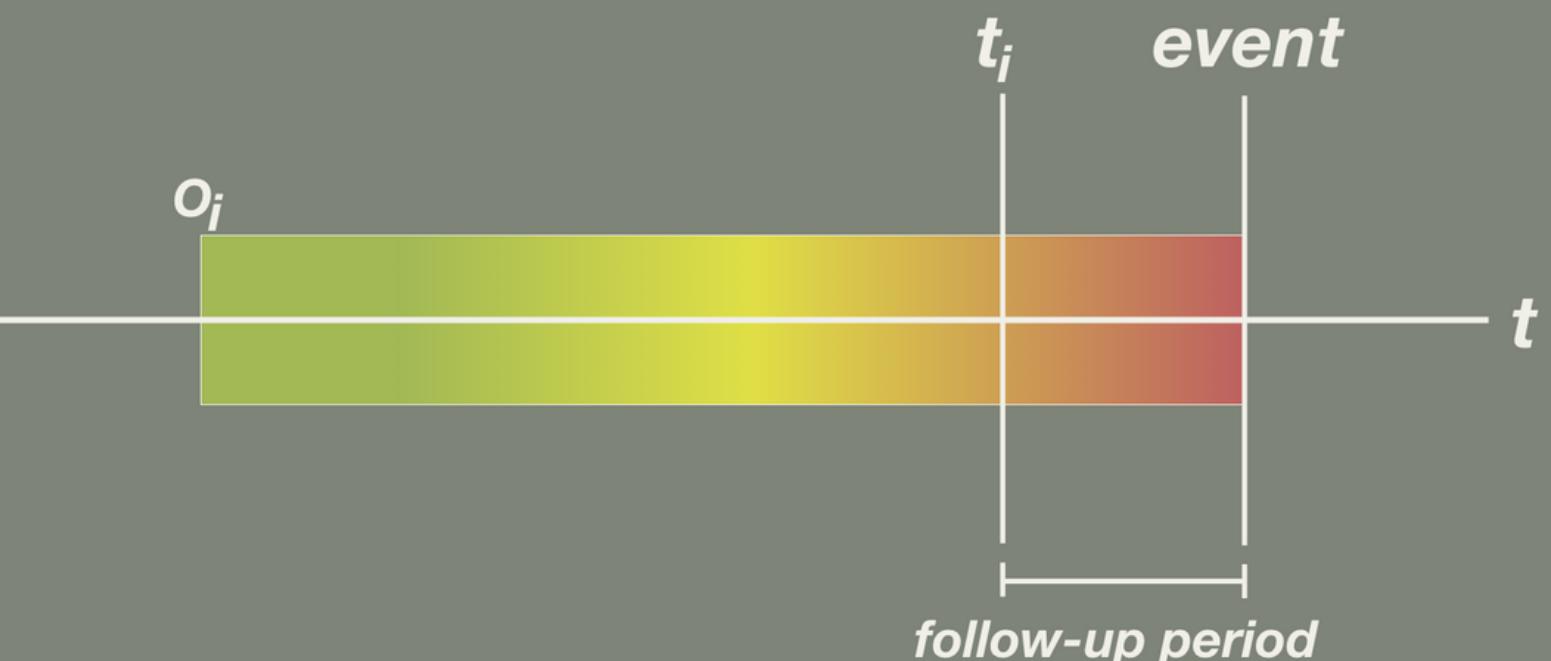
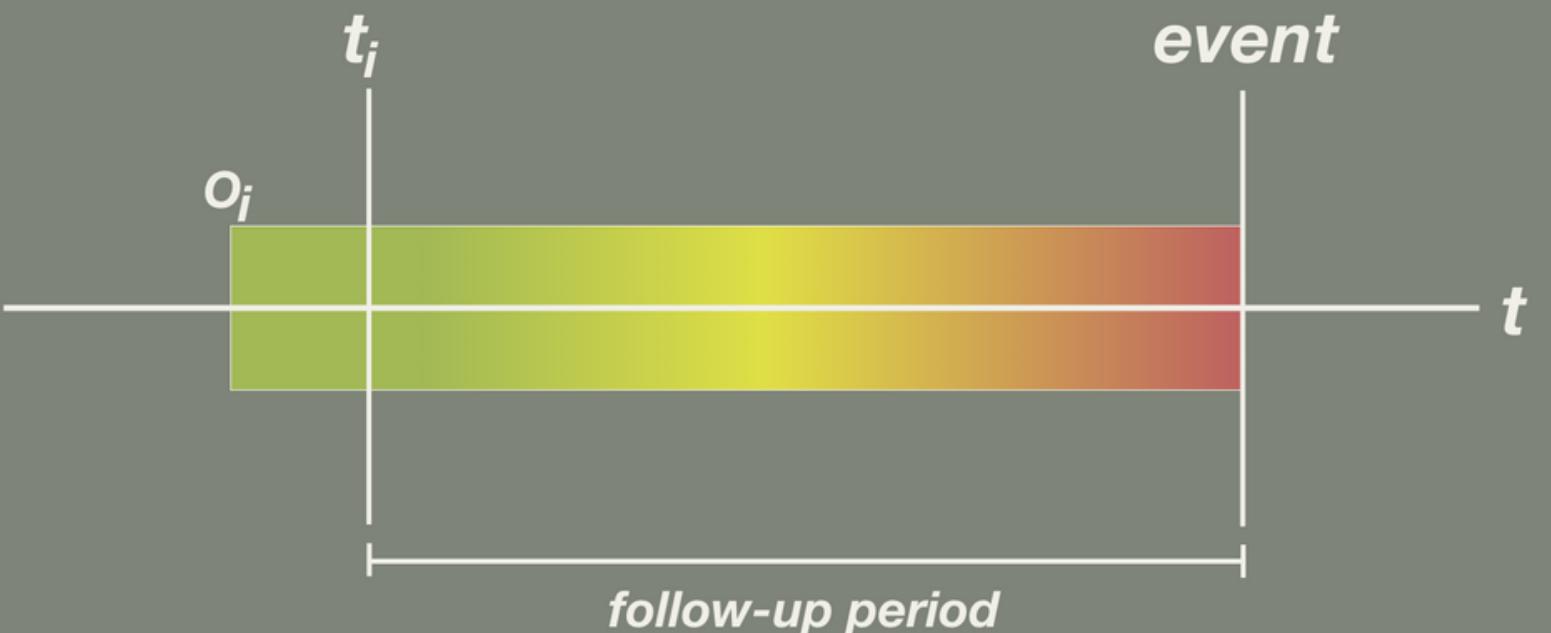


# LEAD-TIME BIAS

Unable to determine exact onset period

Variability in diagnoses time and seriousness of disease

Variations can influences survival and hazard accuracy



# CONCLUSION



Thank You!