



# 9<sup>TH</sup> MALAYSIA STATISTICS CONFERENCE

Department of Statistics, Malaysia

Dealing with Uncertainties: Unearthing Measures for Recovery

## Preliminary study on the survival analysis using censored lung cancer data

Siti Afiqah Muhamad Jamil<sup>1</sup>, Nurain Ibrahim<sup>1</sup>

<sup>1</sup>College of Computing, Informatics and Media, Universiti Teknologi MARA, Shah Alam, 40450, Selangor, Malaysia  
afiqahjamil@uitm.edu.my

### INTRODUCTION

- The primary goal of **Exploratory Data Analysis (EDA)** is to make data "clean," which means it should be free any redundancies.
- The aims of this study are **to perform a preliminary study prior to the application of survival method of analysis on the lung cancer censored observations.**
- This study used the **Kaplan-Meier survival curve, proportional hazard** assumption, **time-varying covariate** assumption by using **Scaled Schoenfeld residuals, Cox-Snell residuals** for the overall goodness of fit of the model assumption, and normality assumption using **quantile-quantile (Q-Q) plot**.
- Despite all survival analysis studies that has been employed previously, the **flow of preliminary study involving survival analysis was not comprehensively explained.**
- Thus, this study could help in **exposing appropriate assumptions of survival analysis.**

Figure 1: Timeline of the duration of survival

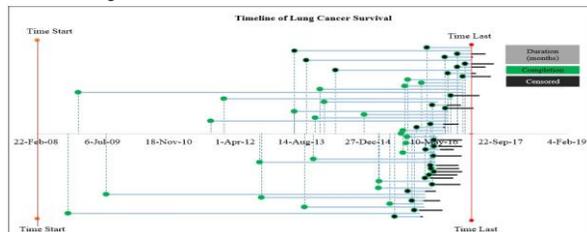
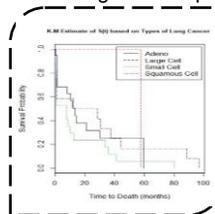


Figure 2: Kaplan Meier survival curve and log-rank test



Log-rank (Mantel-Cox)	z-value	df	P-value
Gender	0.919	1	0.358
Races	-0.289	2	0.773
Types	0.937	3	0.349
Treatments	0.923	3	0.356

Figure 3: Proportional hazard

### METHODOLOGY

**Non-parametric distribution assumption** – Kaplan-Meier (K-M) survival plot and the log-rank test for the categorical variables

**Parametric distribution assumption** – Check on the normality are quantile-quantile (Q-Q) plot or probability (P-P) plot

**Semi-parametric distribution assumption** – Cox proportional hazard model where this study had to test on the proportional hazard assumption and the time varying covariate assumption by using either Scaled Schoenfeld residuals or Unscaled Schoenfeld residuals and either Nelson-Aalen plot or log minus log graph.

Figure 4: Schoenfeld Residuals

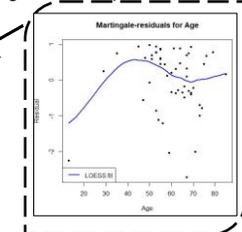
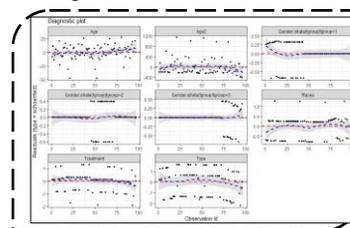
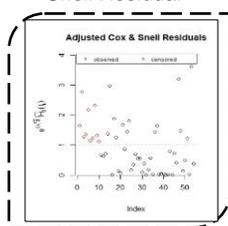
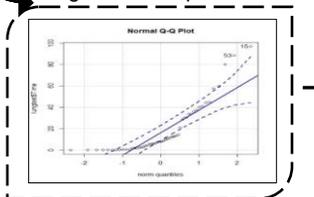


Figure 6: Adjusted Cox-Snell Residual

Figure 5: Q-Q plot



### RESULTS

Table 1: Descriptive statistic for continuous data

Characteristics (Continuous data)	No. (%)
Age (year)	
Mean, Standard deviation (SD)	59.85, 12.30
Median, range	61.50, (13,85)
≤ 60	26 (41.15)
> 60	28 (51.85)

Table 2: Descriptive statistic for categorical data

Characteristics (Categorical data)	No. (%)
Types of lung cancer	
Adenocarcinoma, NSCLC	19 (35.19)
Large Cell Carcinoma, NSCLC	17 (31.48)
Small Cell lung cancer, SCLC	12 (22.22)
Squamous cell carcinoma, NSCLC	6 (11.11)
Treatment of lung cancer	
Chemoradiotherapy, CCRT	32 (59.26)
Chemotherapy, Chemo	5 (9.26)
Chemotherapy and Surgery, Chemosurgery	11 (20.37)
Chemotherapy and Targeted Therapy, ChemoTarget	5 (9.26)

### CONCLUSIONS

- Kaplan-Meier survival plot with log rank test was violated.
- Proportional hazard assumption was violated at one of the categorical variables of gender. The violation of proportional hazard proved that the covariates was not time-fixed and has been solve by using the split method.
- The parametric assumption was satisfied as the quantile-quantile plot shows the constant pattern about 45°. Some outliers also have been detected.
- Overall fit model using Cox-Snell residuals shows that the distribution fit the lung cancer data and cox model also quite good at handling the time-varying covariates.

Since, the assumptions for **semi-parametric cox-proportional hazard and non-parametric Kaplan-Meier of analysis have been violated** which leads to the application of **parametric methods of survival analysis**

### REFERENCES

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