Day9 exercise solutions

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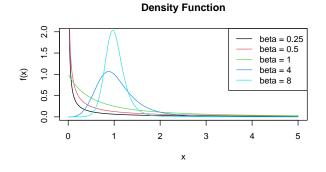
Nov. 15th, 2024

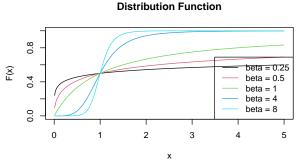
Problem 1: Log-logistic distribution

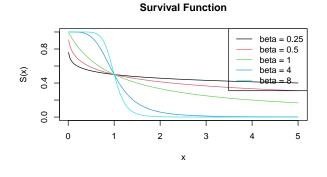
(a) Plot density, distribution, survival, and hazard functions

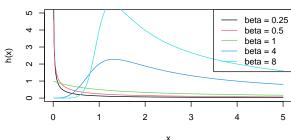
```
# Define the functions
loglogistic_density <- function(x, alpha = 1, beta) {</pre>
  (beta/alpha) * (x/alpha)^(beta-1) / (1 + (x/alpha)^beta)^2
}
loglogistic_distribution <- function(x, alpha = 1, beta) {</pre>
  (x^beta) / (alpha^beta + x^beta)
loglogistic_survival <- function(x, alpha = 1, beta) {</pre>
 1 - loglogistic_distribution(x, alpha, beta)
}
loglogistic_hazard <- function(x, alpha = 1, beta) {</pre>
  loglogistic_density(x, alpha, beta) / loglogistic_survival(x, alpha, beta)
# Create x values
x \leftarrow seq(0.01, 5, length.out = 1000)
beta_values \leftarrow c(0.25, 0.5, 1, 4, 8)
# Create plots
par(mfrow = c(2, 2))
# Density plot
plot(x, loglogistic_density(x, beta = beta_values[1]), type = "l",
     ylim = c(0, 2), main = "Density Function",
     xlab = "x", ylab = "f(x)")
for(beta in beta_values[-1]) {
  lines(x, loglogistic_density(x, beta = beta), col = which(beta_values == beta))
legend("topright", legend = paste("beta =", beta_values),
       col = 1:length(beta_values), lty = 1)
# Distribution plot
plot(x, loglogistic_distribution(x, beta = beta_values[1]), type = "l",
```

```
ylim = c(0, 1), main = "Distribution Function",
     xlab = "x", ylab = "F(x)")
for(beta in beta_values[-1]) {
 lines(x, loglogistic_distribution(x, beta = beta), col = which(beta_values == beta))
legend("bottomright", legend = paste("beta =", beta_values),
       col = 1:length(beta_values), lty = 1)
# Survival plot
plot(x, loglogistic_survival(x, beta = beta_values[1]), type = "l",
     ylim = c(0, 1), main = "Survival Function",
    xlab = "x", ylab = "S(x)")
for(beta in beta values[-1]) {
 lines(x, loglogistic_survival(x, beta = beta), col = which(beta_values == beta))
legend("topright", legend = paste("beta =", beta_values),
       col = 1:length(beta_values), lty = 1)
# Hazard plot
plot(x, loglogistic_hazard(x, beta = beta_values[1]), type = "l",
     ylim = c(0, 5), main = "Hazard Function",
     xlab = "x", ylab = "h(x)")
for(beta in beta_values[-1]) {
  lines(x, loglogistic_hazard(x, beta = beta), col = which(beta_values == beta))
legend("topright", legend = paste("beta =", beta_values),
       col = 1:length(beta values), lty = 1)
```









Hazard Function

```
# Create summary table for log-logistic distribution characteristics
log_logistic_chars <- data.frame(</pre>
  "Parameter" = c("beta < 1", "beta = 1", "beta > 1"),
  "Density Shape" = c("Right-skewed with high peak near zero",
                     "Balanced shape",
                     "More symmetric, bell-shaped"),
  "Hazard Behavior" = c("Decreasing hazard",
                       "Constant hazard",
                       "Initially increasing then decreasing")
)
kable(log_logistic_chars,
      caption = "Log-logistic Distribution Characteristics",
      align = c('l', 'l', 'l'),
      booktabs = TRUE) %>%
  kable_styling(latex_options = c("striped", "hold_position"),
                full_width = FALSE)
```

Table 1: Log-logistic Distribution Characteristics

Parameter	Density.Shape	Hazard.Behavior
beta < 1	Right-skewed with high peak near zero	Decreasing hazard
beta = 1	Balanced shape	Constant hazard
beta > 1	More symmetric, bell-shaped	Initially increasing then decreasing

(b) Weibull Distribution Analysis

```
# Define Weibull density function
weibull_density <- function(x, sigma = 1, kappa) {
    (kappa/sigma) * (x/sigma)^(kappa-1) * exp(-(x/sigma)^kappa)
}

# Create plots for different kappa values
x <- seq(0.01, 5, length.out = 1000)
kappa_values <- c(0.5, 1, 2) # One from each case

par(mfrow = c(1,1))
plot(x, weibull_density(x, kappa = kappa_values[1]), type = "l",
    ylim = c(0, 1.5), main = "Weibull Density for Different Kappa Values",
    xlab = "x", ylab = "f(x)")
for(kappa in kappa_values[-1]) {
    lines(x, weibull_density(x, kappa = kappa), col = which(kappa_values == kappa))
}
legend("topright", legend = paste("kappa =", kappa_values),
    col = 1:length(kappa_values), lty = 1)</pre>
```

Weibull Density for Different Kappa Values

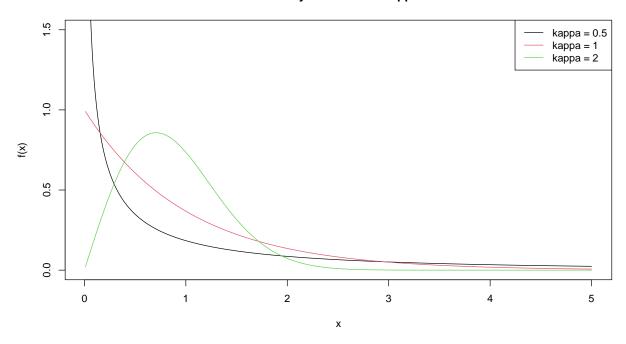


Table 2: Weibull Distribution Characteristics (sigma = 1)

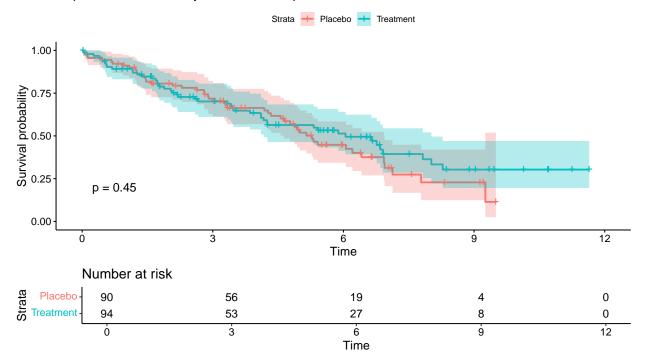
Parameter.Range	Characteristics
0 < kappa < 1	Monotonically decreasing, infinite peak at $x = 0$
kappa = 1	Reduces to exponential distribution
kappa > 1	Peak away from zero, right-skewed

Problem 2: Survival Analysis of PBC Data

Kaplan-Meier Analysis

```
# Plot Kaplan-Meier curves
par(mfrow = c(1,1))
ggsurvplot(km_fit,
data= pbc_data,
pval= TRUE,
conf.int=TRUE,
risk.table= TRUE,
title= "Kaplan-MeierCurvesbyTreatmentGroup",
legend.labs= c("Placebo", "Treatment"))
```

Kaplan-MeierCurvesbyTreatmentGroup



Statistical Analysis

```
# Fit models
cox_model1 <- coxph(surv_obj ~ treat, data = pbc_data)</pre>
cox_model2 <- coxph(surv_obj ~ treat + age + cenc0, data = pbc_data)</pre>
log_rank <- survdiff(surv_obj ~ treat, data = pbc_data)</pre>
test.ph1 <- cox.zph(cox_model1)</pre>
test.ph2 <- cox.zph(cox_model2)</pre>
# Function to create formatted Cox model results
create_cox_table <- function(model) {</pre>
  coef_table <- summary(model)$coefficients</pre>
  conf_table <- summary(model)$conf.int</pre>
 results_df <- data.frame(</pre>
    Variable = rownames(coef_table),
    "Hazard Ratio" = sprintf("%.2f", conf_table[, "exp(coef)"]),
    "95% CI" = sprintf("(%.2f, %.2f)", conf_table[, "lower .95"], conf_table[, "upper .95"]),
    "P-value" = sprintf("%.3f", coef_table[, "Pr(>|z|)"]),
    check.names = FALSE
  )
  concordance <- sprintf("%.3f", summary(model)$concordance[1])</pre>
 kable(results df,
        caption = paste("Cox Model Results (Concordance =", concordance, ")"),
        align = c('l', 'c', 'c', 'c'),
        booktabs = TRUE) %>%
    kable_styling(latex_options = c("striped", "hold_position"),
                full_width = FALSE)
}
# Basic Cox Model Results
cat("\n### Basic Cox Model\n")
```

Basic Cox Model

```
create_cox_table(cox_model1)
```

Table 3: Cox Model Results (Concordance = 0.503)

Variable	Hazard Ratio	95% CI	P-value
treatTreatment	0.86	(0.57, 1.28)	0.451

```
# Extended Cox Model Results
cat("\n### Extended Cox Model\n")
```

Extended Cox Model

```
create_cox_table(cox_model2)
```

Table 4: Cox Model Results (Concordance = 0.669)

Variable	Hazard Ratio	95% CI	P-value
treatTreatment	0.83	(0.55, 1.24)	0.362
age	1.04	(1.02, 1.06)	0.001
cenc0	3.85	(2.38, 6.23)	0.000

```
# Log-rank test results
cat("\n### Log-rank Test Results\n")
```

Log-rank Test Results

```
logrank_df <- data.frame(</pre>
  Group = c("Placebo", "Treatment"),
  Observed = c(log_rank$obs[1], log_rank$obs[2]),
  Expected = c(log_rank$exp[1], log_rank$exp[2]),
  "(0-E)^2/E" = c(
    (log_rank$obs[1] - log_rank$exp[1])^2/log_rank$exp[1],
    (log_rank$obs[2] - log_rank$exp[2])^2/log_rank$exp[2]
  )
)
kable(logrank_df,
      caption = sprintf("Log-rank Test Results (p-value = %.3f)",
                       1 - pchisq(log_rank$chisq, df = 1)),
      align = c('l', 'c', 'c', 'c'),
      digits = 2,
      booktabs = TRUE) %>%
  kable_styling(latex_options = c("striped", "hold_position"),
                full width = FALSE)
```

Table 5: Log-rank Test Results (p-value = 0.450)

Group	Observed	Expected	X.O.E2.E
Placebo	49	45.33	0.30
Treatment	47	50.67	0.27

```
# Proportional hazards test results
cat("\n### Proportional Hazards Test Results\n")
```

Proportional Hazards Test Results

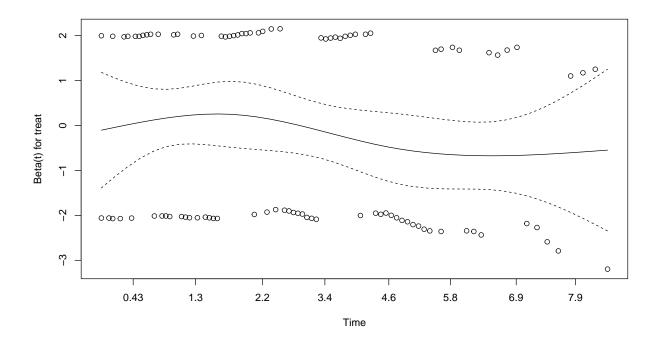
Table 6: Proportional Hazards Assumption Test

Variable	Chi-square	P-value
treat	1.04	0.308
age	0.04	0.839
cenc0	0.26	0.607
GLOBAL	1.33	0.722

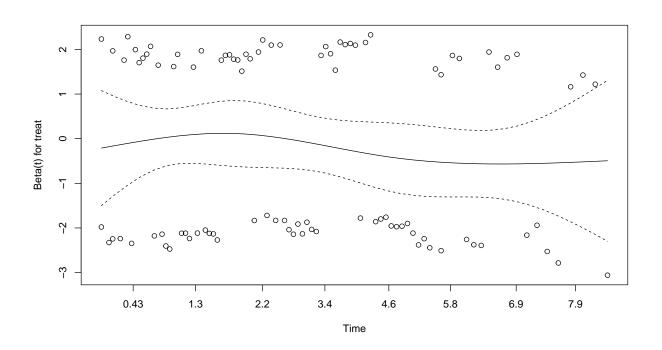
```
# Key findings summary
cat("\n### Key Findings\n")
```

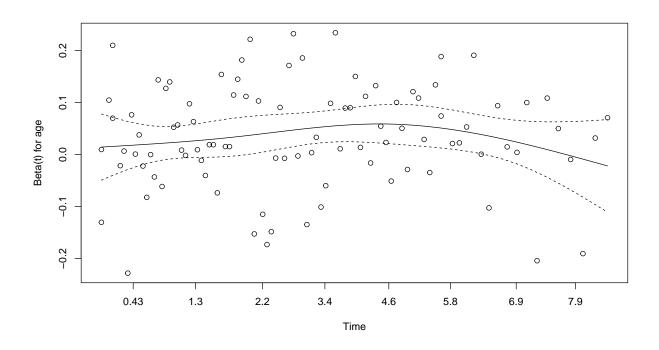
Key Findings

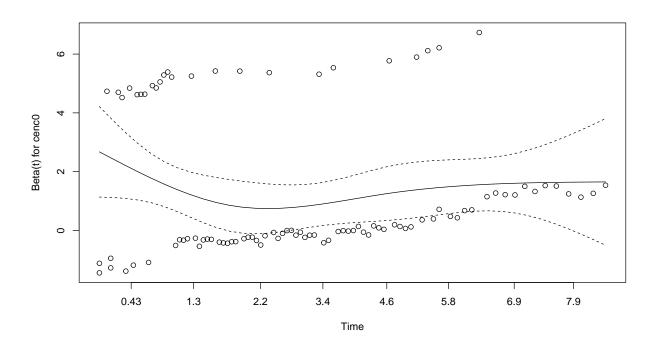
```
findings_df <- data.frame(
   Category = c("Treatment Effect", "Risk Factors", "Model Performance", "Assumptions"),
   Finding = c(
    "17% reduction in hazard (non-significant, p > 0.05)",
    "Age: 4% increased risk per year (p < 0.001)\nCentral cholestasis: 284% increased risk (p < 0.001)"
   "Extended model concordance: 0.669\nSignificant improvement over basic model",
   "Proportional hazards satisfied for all variables (p > 0.05)"
  )
  plot(test.ph1)
```



plot(test.ph2)







full_width = FALSE)

Table 7: Summary of Key Findings

Category	Details
Treatment Effect	17% reduction in hazard (non-significant, p > 0.05)
Risk Factors	Age: 4% increased risk per year (p < 0.001) Central cholestasis: 284% increased risk (p < 0.001)
Model Performance	Extended model concordance: 0.669 Significant improvement over basic model
Assumptions	Proportional hazards satisfied for all variables $(p > 0.05)$