## Simulation\_data\_generate

```
1. Choose different cox distributions to generate data
  2. Choose different beta
  3. Create functions to simplify codes
  4. Visualization 4.1 KM-curve vs. fitted curve 4.2 beta ~ MSE with different parameters to generate data
  5. discussion: description, pros and cons ...
## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5
                      v purrr
                                0.3.4
## v tibble 3.1.4
                      v dplyr
                                1.0.7
## v tidyr
           1.1.3
                      v stringr 1.4.0
           2.0.1
## v readr
                      v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## Attaching package: 'pracma'
## The following object is masked from 'package:purrr':
##
##
      cross
## Loading required package: ggpubr
##
## Attaching package: 'survminer'
## The following object is masked from 'package:survival':
```

## Generate survival data

myeloma

##

```
set.seed(666)

# Generate survival data
genn_dat <- function(n, lambda, beta, gamma, alpha, dist) {
    # Generate key for each observation
    id <- seq(1:n)

# Predictor X (treatment = 1; control = 0)
x <- rbinom(n, size = 1, prob = 0.5)

## --- Generate Survival Time T ---
U <- runif(n)
# Use exponential distribution
expo <- function(U, lambda, x, beta) {</pre>
```

```
-\log(U) / (lambda * exp(x * beta))
  # Use weibull distribution
  weibull <- function(U, lambda, x, gamma, beta) {</pre>
    (-\log(U) / (lambda * exp(x * beta))) ^ (1 / gamma)
  # --- To be modified (add cox function) ---
  # Use 'Cox' distribution -- gompertz
  gompertz <- function(U, alpha, lambda, x, beta) {</pre>
    (1 / alpha) * (1 - alpha * log(U) / (lambda * exp(x * beta)))
  # Select different baseline functions
  if (dist == "expo") {
    surv_t <- expo(U, lambda, x, beta)}</pre>
  else if (dist == "weibull") {
    surv_t <- weibull(U, lambda, x, gamma, beta)</pre>
  }
  else {
    surv_t <- gompertz(U, alpha, lambda, x, beta)</pre>
 return(
   df = data.frame(
     id = id,
     treatment = x,
      time = surv_t
    )
 )
}
```

## Visualization(beta = 0.5)

```
par(oma = c(3,3,3,3))
par(mar = c(2,1,1,1))

### 1. data-exponential(lambda = 2)
## Generate Data
set.seed(666)
dat_1 <- genn_dat(n = 100, lambda = 2, beta = 0.5, dist = "expo")
s_1 <- with(dat_1, Surv(time))

par(mfrow=c(5,3))
## Exponential
fit_expo_1 = flexsurvreg(s_1 ~ as.factor(treatment), dist = 'exponential', data = dat_1)
plot(fit_expo_1, col = c('red', 'blue'), main = "Exponential", ylab = "Exponential(2)",cex.lab=1.5, fon legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)

## Weibull
fit_wei_1 = flexsurvreg(s_1 ~ as.factor(treatment), dist = 'weibull', data = dat_1)
plot(fit_wei_1, col = c('red', 'blue'), main = "Weibull")</pre>
```

```
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Cox
fit_km_1 = survfit(s_1 ~ treatment, data = dat_1)
plot(fit_km_1, conf.int = F, main = "Cox")
fit_cox_1 = coxph(s_1 ~ as.factor(treatment), data = dat_1)
lines(survfit(fit_cox_1, newdata = data.frame(treatment = 0)), col = "blue", conf.int = F)
lines(survfit(fit cox 1, newdata = data.frame(treatment = 1)), col = "red", conf.int = F)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
### 2. data-weibull(lambda = 2, gamma = 1)
## Generate Data
set.seed(666)
dat_2 \leftarrow genn_dat(n = 100, lambda = 2, beta = 0.5, gamma = 1, dist = "weibull")
s_2 = with(dat_2, Surv(time))
## Exponential
fit_expo_2 = flexsurvreg(s_2 ~ as.factor(treatment), dist = 'exponential', data = dat_2)
plot(fit_expo_2, col = c('red', 'blue'), main = "Exponential", ylab = "Weibull(1)", cex.lab=1.5, font.l
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Weibull
fit_wei_2 = flexsurvreg(s_2 ~ as.factor(treatment), dist = 'weibull', data = dat_2)
plot(fit_wei_2, col = c('red', 'blue'), main = "Weibull")
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Cox
fit_km_2 = survfit(s_2 ~ treatment, data = dat_2)
plot(fit_km_2, conf.int = F, main = "Cox")
fit_cox_2 = coxph(s_2 ~ as.factor(treatment), data = dat_2)
lines(survfit(fit_cox_2, newdata = data.frame(treatment = 0)), col = "blue", conf.int = F)
lines(survfit(fit_cox_2, newdata = data.frame(treatment = 1)), col = "red", conf.int = F)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
### 3. data-weibull(lambda = 2, gamma = 3)
## Generate Data
set.seed(666)
dat_4 <- genn_dat(n = 100, lambda = 2, beta = 0.5, gamma = 3, dist = "weibull")
s_4 = with(dat_4, Surv(time))
## Exponential
fit_expo_4 = flexsurvreg(s_4 ~ as.factor(treatment), dist = 'exponential', data = dat_4)
plot(fit_expo_4, col = c('red', 'blue'), main = "Exponential", ylab = "Weibull(3)", cex.lab=1.5, font.l
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Weibull
fit_wei_4 = flexsurvreg(s_4 ~ as.factor(treatment), dist = 'weibull', data = dat_4)
plot(fit_wei_4, col = c('red', 'blue'), main = "Weibull")
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Cox
fit_km_4 = survfit(s_4 ~ treatment, data = dat_4)
plot(fit_km_4, conf.int = F, main = "Cox")
```

```
fit_cox_4 = coxph(s_4 \sim as.factor(treatment), data = dat_4)
lines(survfit(fit_cox_4, newdata = data.frame(treatment = 0)), col = "blue", conf.int = F)
lines(survfit(fit_cox_4, newdata = data.frame(treatment = 1)), col = "red", conf.int = F)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
### 4. data-gompertz(alpha = 2, lambda = 2)
## Generate Data
set.seed(666)
dat_3 <- genn_dat(n = 100, lambda = 2, beta = 0.5, alpha = 2, dist = "gompertz")</pre>
s 3 = with(dat 3, Surv(time))
## Exponential
fit_expo_3 = flexsurvreg(s_3 ~ as.factor(treatment), dist = 'exponential', data = dat_3)
plot(fit_expo_3, col = c('red', 'blue'), main = "Exponential", ylab = "Gompertz(2)", cex.lab=1.5, font.
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Weibull
fit_wei_3 = flexsurvreg(s_3 ~ as.factor(treatment), dist = 'weibull', data = dat_3)
plot(fit_wei_3, col = c('red', 'blue'), main = "Weibull", cex.lab=1.5)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
fit_km = survfit(s_3 ~ treatment, data = dat_3)
plot(fit_km, conf.int = F, main = "Cox", cex.lab=1.5)
fit_cox_3 = coxph(s_3 ~ as.factor(treatment), data = dat_3)
lines(survfit(fit_cox_3, newdata = data.frame(treatment = 0)), col = "blue", conf.int = F)
lines(survfit(fit_cox_3, newdata = data.frame(treatment = 1)), col = "red", conf.int = F)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
### 5. data-gompertz(lambda = 2, alpha = 4)
## Generate Data
set.seed(666)
dat_5 \leftarrow genn_dat(n = 100, lambda = 2, beta = 0.5, alpha = 4, dist = "gompertz")
s_5 = with(dat_5, Surv(time))
## Exponential
fit_expo_5 = flexsurvreg(s_5 ~ as.factor(treatment), dist = 'exponential', data = dat_5)
plot(fit_expo_5, col = c('red', 'blue'), main = "Exponential", ylab = "Gompertz(4)", xlab = "t", cex.la
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Weibull
fit_wei_5 = flexsurvreg(s_5 ~ as.factor(treatment), dist = 'weibull', data = dat_5)
plot(fit_wei_5, col = c('red', 'blue'), main = "Weibull", xlab = "t", cex.lab=1.5)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
## Cox
fit_km = survfit(s_5 ~ treatment, data = dat_5)
plot(fit_km, conf.int = F, main = "Cox", xlab = "t", cex.lab=1.5)
fit_cox_5 = coxph(s_5 \sim as.factor(treatment), data = dat_5)
lines(survfit(fit_cox_5, newdata = data.frame(treatment = 0)), col = "blue", conf.int = F)
lines(survfit(fit_cox_5, newdata = data.frame(treatment = 1)), col = "red", conf.int = F)
legend("topright", legend = c('Treatment', 'Control'), col = c('red', 'blue'), lty = 1, cex = 0.8)
```

```
Exponential
                                       Weibull
                                                                    Cox
                                                                            Treatment
                 2.0
                                      1.0
                                            2.0
                                                                 1.0
                                                                        2.0
   0.0
          1.0
                       3.0
                               0.0
                                                   3.0
                                                          0.0
                                                                              3.0
                                                                    Cox
         Exponential
                                       Weibull
                    Treatment
                                                Treatment O
                                                                            Treatment
                 2.0
                       3.0
                                      1.0
                                            2.0
                                                                        2.0
   0.0
          1.0
                               0.0
                                                   3.0
                                                          0.0
                                                                 1.0
                                                                               3.0
         Exponential
                                       Weibull
                                                                    Cox
                                                Treatment O
                        ment 0
                                                                            Treatment
                                      0.5
   0.0
          0.5
                  1.0
                         1.5
                               0.0
                                             1.0
                                                     1.5
                                                          0.0
                                                                  0.5
                                                                         1.0
                                                                                1.5
                                       Weibull
                                                                    Cox
         Exponential
                    Treatment
                                                Treatment
                                                                            Treatment
                     3
                               0
                                           2
                                                 3
                                                           0
                                                                       2
                                                                             3
                                                                    Cox
         Exponential
                                       Weibull
                                                Treatment O
                                                                            Treatment
         1.0
                2.0
                      3.0
                               0.0
                                     1.0
                                           2.0
                                                  3.0
                                                          0.0
                                                                 1.0
                                                                       2.0
                                                                             3.0
   0.0
3.MSE vs. beta
# Simulation results
sim_result <- function(iteration, n, lambda, beta, gamma, alpha, dist) {</pre>
  # Store estimated beta
  result <- tibble(</pre>
  expo_beta = rep(NA, iteration),
  weibull_beta = rep(NA, iteration),
  cox_beta = rep(NA, iteration)
  )
  for(i in 1:iteration) {
    # Generate data
    data <- genn_dat(n, lambda, beta, gamma, alpha, dist)</pre>
    # Use exponential model
    fit_expo <- survreg(Surv(data$time) ~ data$treatment, dist = "exponential")</pre>
    result$expo_beta[i] <- -fit_expo$coefficients[-1]</pre>
    # Use weibull model
    fit_weibull <- survreg(Surv(data$time) ~ data$treatment, dist = "weibull")</pre>
    result$weibull_beta[i] <- -fit_weibull$coefficients[-1] / fit_weibull$scale
    # Use gompertz model
    fit_cox <- coxph(Surv(data$time) ~ data$treatment)</pre>
    result$cox_beta[i] <- fit_cox$coefficients</pre>
  }
  # Calculate MSE
  return(result)
}
# Under 1000 simulations & 100 sample sizes
# True\ beta = 0.5
# Exponential beta
iteration = 1000
```

```
n = 100
beta = 0.5
expo result <- sim result(iteration, n, lambda = 2, beta, dist = "expo")
expo beta <- tibble(</pre>
 expo_mse = sum((expo_result$expo_beta - beta)^2) / iteration,
 expo_mean = mean(expo_result$expo_beta),
 expo_sd = sd(expo_result$expo_beta),
 weibull_mse = sum((expo_result$weibull_beta - beta)^2) / iteration,
 weibull mean = mean(expo result$weibull beta),
 weibull_sd = sd(expo_result$weibull_beta),
 cox_mse = sum((expo_result$cox_beta - beta)^2) / iteration,
 cox_mean = mean(expo_result$cox_beta),
 cox_sd = sd(expo_result$cox_beta)
# Weibull beta (gamma = 1)
weibull_result_gamma1 <- sim_result(iteration, n , beta, lambda = 2, gamma = 1, dist = "weibull")
weibull_beta_gamma1 <- tibble(</pre>
  expo_mse = sum((weibull_result_gamma1$expo_beta - beta)^2) / iteration,
  expo_mean = mean(weibull_result_gamma1$expo_beta),
  expo sd = sd(weibull result gamma1$expo beta),
  weibull_mse = sum((weibull_result_gamma1$weibull_beta - beta)^2) / iteration,
 weibull_mean = mean(weibull_result_gamma1$weibull_beta),
 weibull_sd = sd(weibull_result_gamma1$weibull_beta),
 cox_mse = sum((weibull_result_gamma1$cox_beta - beta)^2) / iteration,
 cox_mean = mean(weibull_result_gamma1$cox_beta),
  cox_sd = sd(weibull_result_gamma1$cox_beta)
# Weibull beta (gamma = 3)
weibull_result_gamma3 <- sim_result(iteration, n , beta, lambda = 2, gamma = 3, dist = "weibull")
weibull_beta_gamma3 <- tibble(</pre>
  expo_mse = sum((weibull_result_gamma3$expo_beta - beta)^2) / iteration,
  expo_mean = mean(weibull_result_gamma3$expo_beta),
  expo_sd = sd(weibull_result_gamma3$expo_beta),
  weibull_mse = sum((weibull_result_gamma3$weibull_beta - beta)^2) / iteration,
  weibull_mean = mean(weibull_result_gamma3$weibull_beta),
 weibull_sd = sd(weibull_result_gamma3$weibull_beta),
 cox mse = sum((weibull result gamma3$cox beta - beta)^2) / iteration,
 cox_mean = mean(weibull_result_gamma3$cox_beta),
 cox_sd = sd(weibull_result_gamma3$cox_beta)
)
# Gompertz beta (alpha = 2)
Gompertz_result_alpha2 <- sim_result(iteration, n , beta, lambda = 2, alpha = 2, dist = "Gompertz")</pre>
Gompertz_beta_alpha2 <- tibble(</pre>
  expo_mse = sum((Gompertz_result_alpha2$expo_beta - beta)^2) / iteration,
  expo_mean = mean(Gompertz_result_alpha2$expo_beta),
  expo_sd = sd(Gompertz_result_alpha2$expo_beta),
  weibull_mse = sum((Gompertz_result_alpha2$weibull_beta - beta)^2) / iteration,
  weibull_mean = mean(Gompertz_result_alpha2$weibull_beta),
  weibull_sd = sd(Gompertz_result_alpha2$weibull_beta),
  cox_mse = sum((Gompertz_result_alpha2$cox_beta - beta)^2) / iteration,
```

```
cox_mean = mean(Gompertz_result_alpha2$cox_beta),
    cox_sd = sd(Gompertz_result_alpha2$cox_beta)
)

# Gompertz beta (alpha = 4)
Gompertz_result_alpha4 <- sim_result(iteration, n , beta, lambda = 2, alpha = 4, dist = "Gompertz")
Gompertz_beta_alpha4 <- tibble(
    expo_mse = sum((Gompertz_result_alpha4$expo_beta - beta)^2) / iteration,
    expo_mean = mean(Gompertz_result_alpha4$expo_beta),
    expo_sd = sd(Gompertz_result_alpha4$expo_beta),
    weibull_mse = sum((Gompertz_result_alpha4$weibull_beta - beta)^2) / iteration,
    weibull_mean = mean(Gompertz_result_alpha4$weibull_beta),
    cox_mse = sum((Gompertz_result_alpha4$veibull_beta),
    cox_mse = sum((Gompertz_result_alpha4$cox_beta - beta)^2) / iteration,
    cox_mean = mean(Gompertz_result_alpha4$cox_beta)
)</pre>
```