

# CaseStudy

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This R Markdown document includes the codes and figures in Section 4.

```
source("Scales.R")
source("Functions.R")
library(sensitivity)
library(cowplot)
library(ggplot2)
```

## Summary of Core Parameters

N = size of grid  
M = the number of realization  
x.sample = initial input data  
y.sample = initial output data  
g.sample = initial constraint data  
lb = lower bound - vector  
ub = upper bound - vector  
g\_constraint = constraint threshold

Sensitivity Analysis – Run Sobol' Analysis Using Saltelli's Scheme

```
lb <- c(125, 150, 400, 7200)
ub <- c(180, 180, 6600, 12000)
eps <- sqrt(.Machine$double.eps)
g_constraint <- 0.96
d <- 4
initial_sample_size <- 50
N <- 500
M <- 10000
```

Load Cure Process Data

```
load("Q22.Rdata")
x.sample <- Q22[,1:4]
y.sample <- Q22[,5]
g.sample <- Q22[,6]
```

## Decision Uncertainty Experiment: Optimal Decision Uncertainty & Sensativity Analysis (Section 4)

```
set.seed(123)
```

```
result_gp <- gp_opts_constraint(x.sample, y.sample, g.sample, g_constraint, lb, ub, N, M, d)
```

```
##
## optimisation start
## -----
## * estimation method : MLE
## * optimisation method : BFGS
## * analytical gradient : used
## * trend model : 1
## * covariance model :
## - type : matern5.2
## - nugget : 1.490116e-08
## - parameters lower bounds : 1e-10 1e-10 1e-10 1e-10
## - parameters upper bounds : 1.890909 1.933333 1.953333 1.891667
## - variance bounds : 0.03166104 5.123977
## - best initial criterion value(s) : 67.8006
##
## N = 5, M = 5 machine precision = 2.22045e-16
## At X0, 0 variables are exactly at the bounds
## At iterate 0 f = -67.801 [proj g] = 4.8157
## At iterate 1 f = -79.666 [proj g] = 1.3889
## At iterate 2 f = -80.238 [proj g] = 1.3741
## At iterate 3 f = -81.11 [proj g] = 3.5765
## At iterate 4 f = -81.4 [proj g] = 1.5847
## At iterate 5 f = -81.986 [proj g] = 1.5424
## At iterate 6 f = -82.299 [proj g] = 1.5037
## At iterate 7 f = -83.257 [proj g] = 4.8287
## At iterate 8 f = -83.386 [proj g] = 1.4822
## At iterate 9 f = -83.52 [proj g] = 0.55005
## At iterate 10 f = -83.522 [proj g] = 0.13967
## At iterate 11 f = -83.523 [proj g] = 0.17992
## At iterate 12 f = -83.523 [proj g] = 0.08929
## At iterate 13 f = -83.523 [proj g] = 0.083954
## At iterate 14 f = -83.523 [proj g] = 0.015509
## At iterate 15 f = -83.523 [proj g] = 0.00081121
## At iterate 16 f = -83.523 [proj g] = 0.00017932
##
## iterations 16
## function evaluations 20
## segments explored during Cauchy searches 20
## BFGS updates skipped 0
## active bounds at final generalized Cauchy point 1
## norm of the final projected gradient 0.000179315
## final function value -83.5232
##
## F = -83.5232
## final value -83.523162
## converged
##
## optimisation start
## -----
## * estimation method : MLE
## * optimisation method : BFGS
## * analytical gradient : used
## * trend model : 1
## * covariance model :
## - type : matern5.2
## - nugget : 1.490116e-08
## - parameters lower bounds : 1e-10 1e-10 1e-10 1e-10
## - parameters upper bounds : 1.890909 1.933333 1.953333 1.891667
## - variance bounds : 3.054316e-05 0.00587365
## - best initial criterion value(s) : 209.7215
##
## N = 5, M = 5 machine precision = 2.22045e-16
## At X0, 0 variables are exactly at the bounds
## At iterate 0 f = -209.72 [proj g] = 1.6862
## At iterate 1 f = -216.9 [proj g] = 1.6767
## At iterate 2 f = -219.25 [proj g] = 1.6652
## At iterate 3 f = -221.61 [proj g] = 1.6609
## At iterate 4 f = -221.83 [proj g] = 1.6617
## At iterate 5 f = -221.86 [proj g] = 1.6639
## At iterate 6 f = -221.91 [proj g] = 1.6698
## At iterate 7 f = -221.95 [proj g] = 1.6766
## At iterate 8 f = -221.95 [proj g] = 1.677
## At iterate 9 f = -221.95 [proj g] = 1.6771
## At iterate 10 f = -221.96 [proj g] = 1.6771
## At iterate 11 f = -221.96 [proj g] = 1.677
## At iterate 12 f = -221.96 [proj g] = 1.6765
## At iterate 13 f = -221.98 [proj g] = 1.6765
## At iterate 14 f = -222.03 [proj g] = 1.6749
## At iterate 15 f = -222.14 [proj g] = 1.6706
## At iterate 16 f = -222.4 [proj g] = 1.6604
## At iterate 17 f = -222.93 [proj g] = 1.6401
## At iterate 18 f = -224.11 [proj g] = 1.6056
## At iterate 19 f = -225.33 [proj g] = 1.5847
## At iterate 20 f = -225.65 [proj g] = 1.5643
## At iterate 21 f = -225.73 [proj g] = 1.5502
## At iterate 22 f = -225.75 [proj g] = 0.90517
## At iterate 23 f = -225.84 [proj g] = 0.88894
## At iterate 24 f = -225.95 [proj g] = 0.87733
## At iterate 25 f = -225.95 [proj g] = 0.87458
## At iterate 26 f = -226.01 [proj g] = 0.88388
## At iterate 27 f = -226.01 [proj g] = 0.88851
## At iterate 28 f = -226.03 [proj g] = 0.89357
## At iterate 29 f = -226.04 [proj g] = 0.88973
## At iterate 30 f = -226.05 [proj g] = 0.88551
## At iterate 31 f = -226.05 [proj g] = 0.88626
## At iterate 32 f = -226.06 [proj g] = 0.87971
## At iterate 33 f = -226.06 [proj g] = 0.88081
## At iterate 34 f = -226.07 [proj g] = 0.87088
## At iterate 35 f = -226.23 [proj g] = 0.82594
## At iterate 36 f = -228.43 [proj g] = 1.6078
## At iterate 37 f = -228.43 [proj g] = 1.6069
## At iterate 38 f = -230.22 [proj g] = 1.6277
## At iterate 39 f = -230.23 [proj g] = 1.6277
## At iterate 40 f = -230.78 [proj g] = 0.45868
## At iterate 41 f = -230.98 [proj g] = 0.68333
## At iterate 42 f = -230.98 [proj g] = 0.45757
## At iterate 43 f = -230.98 [proj g] = 0.45756
## At iterate 44 f = -230.98 [proj g] = 0.45756
## At iterate 45 f = -230.98 [proj g] = 0.45756
##
## iterations 45
## function evaluations 62
## segments explored during Cauchy searches 50
## BFGS updates skipped 0
## active bounds at final generalized Cauchy point 2
## norm of the final projected gradient 0.457563
## final function value -230.978
##
## F = -230.978
## final value -230.977767
## converged
```

```
x_gp <- result_gp$gp.xhat
y_gp <- result_gp$gp.yhat
g_gp <- result_gp$gp.ghat
surrogate_y <- function(X){result_gp$gp.y(X)$mean}
surrogate_g <- function(X){result_gp$gp.g(X)$mean}

opt_mean <- gp_opt_mean_constraint(x.sample, y.sample, g.sample, g_constraint, lb, ub)
```

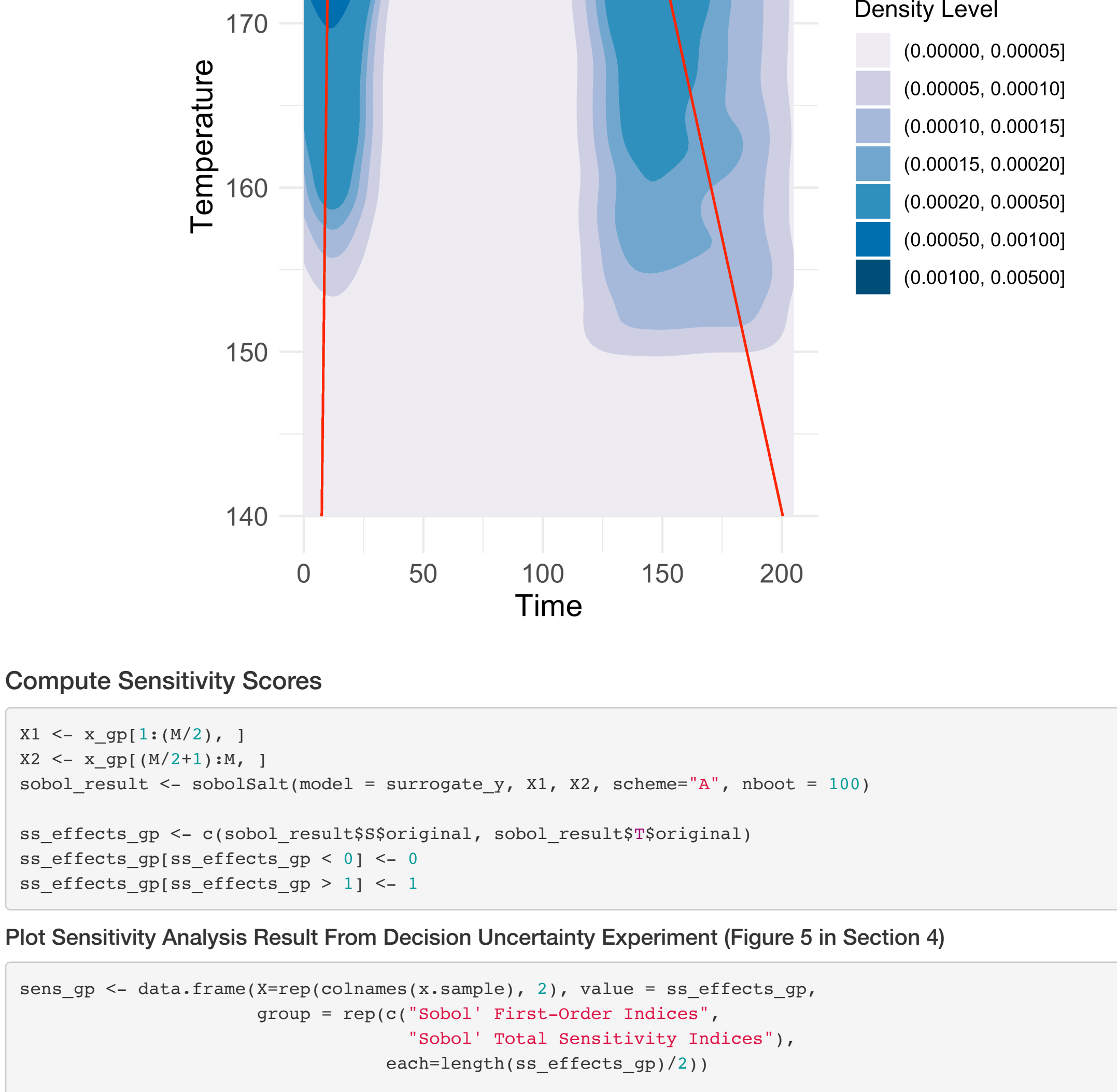
```
##
## optimisation start
## -----
## * estimation method : MLE
## * optimisation method : BFGS
## * analytical gradient : used
## * trend model : 1
## * covariance model :
## - type : matern5.2
## - nugget : 1.490116e-08
## - parameters lower bounds : 1e-10 1e-10 1e-10 1e-10
## - parameters upper bounds : 1.890909 1.933333 1.953333 1.891667
## - variance bounds : 0.03166104 5.123977
## - best initial criterion value(s) : 76.92061
##
## N = 5, M = 5 machine precision = 2.22045e-16
## At X0, 0 variables are exactly at the bounds
## At iterate 0 f = -76.921 [proj g] = 4.8639
## At iterate 1 f = -77.459 [proj g] = 1.4312
## At iterate 2 f = -78.207 [proj g] = 1.395
## At iterate 3 f = -79.496 [proj g] = 4.726
## At iterate 4 f = -80.008 [proj g] = 4.634
## At iterate 5 f = -82.406 [proj g] = 1.3621
## At iterate 6 f = -83.158 [proj g] = 1.4797
## At iterate 7 f = -83.33 [proj g] = 1.457
## At iterate 8 f = -83.424 [proj g] = 1.2689
## At iterate 9 f = -83.452 [proj g] = 1.3648
## At iterate 10 f = -83.49 [proj g] = 1.3436
## At iterate 11 f = -83.523 [proj g] = 0.3711
## At iterate 12 f = -83.523 [proj g] = 0.11971
## At iterate 13 f = -83.523 [proj g] = 0.019497
## At iterate 14 f = -83.523 [proj g] = 0.018109
##
## iterations 14
## function evaluations 18
## segments explored during Cauchy searches 18
## BFGS updates skipped 0
## active bounds at final generalized Cauchy point 1
## norm of the final projected gradient 0.0181093
## final function value -83.5232
##
## F = -83.5232
## final value -83.523161
## converged
##
## optimisation start
## -----
## * estimation method : MLE
## * optimisation method : BFGS
## * analytical gradient : used
## * trend model : 1
## * covariance model :
## - type : matern5.2
## - nugget : 1.490116e-08
## - parameters lower bounds : 1e-10 1e-10 1e-10 1e-10
## - parameters upper bounds : 1.890909 1.933333 1.953333 1.891667
## - variance bounds : 3.054316e-05 0.00587365
## - best initial criterion value(s) : 216.0585
##
## N = 5, M = 5 machine precision = 2.22045e-16
## At X0, 0 variables are exactly at the bounds
## At iterate 0 f = -216.06 [proj g] = 1.7357
## At iterate 1 f = -224.11 [proj g] = 1.446
## At iterate 2 f = -225.51 [proj g] = 0.8018
## At iterate 3 f = -225.64 [proj g] = 0.81554
## At iterate 4 f = -225.69 [proj g] = 0.83361
## At iterate 5 f = -225.69 [proj g] = 1.1663
## At iterate 6 f = -225.7 [proj g] = 0.82666
## At iterate 7 f = -225.7 [proj g] = 0.83073
## At iterate 8 f = -225.7 [proj g] = 0.83056
## At iterate 9 f = -225.71 [proj g] = 0.82893
## At iterate 10 f = -225.72 [proj g] = 0.82452
## At iterate 11 f = -225.74 [proj g] = 1.4608
## At iterate 12 f = -225.8 [proj g] = 1.4623
## At iterate 13 f = -227.07 [proj g] = 1.4825
## At iterate 14 f = -227.27 [proj g] = 1.5297
## At iterate 15 f = -229.67 [proj g] = 1.4881
## At iterate 16 f = -230.05 [proj g] = 0.62612
## At iterate 17 f = -230.05 [proj g] = 0.87658
## At iterate 18 f = -230.06 [proj g] = 1.3249
## At iterate 19 f = -230.09 [proj g] = 1.4858
## At iterate 20 f = -230.16 [proj g] = 1.4808
## At iterate 21 f = -230.26 [proj g] = 1.472
## At iterate 22 f = -230.38 [proj g] = 1.5808
## At iterate 23 f = -230.44 [proj g] = 0.8629
## At iterate 24 f = -230.44 [proj g] = 1.2415
## At iterate 25 f = -230.44 [proj g] = 1.1408
## At iterate 26 f = -230.44 [proj g] = 1.1367
##
## iterations 26
## function evaluations 35
## segments explored during Cauchy searches 32
## BFGS updates skipped 0
## active bounds at final generalized Cauchy point 1
## norm of the final projected gradient 1.13672
## final function value -230.441
##
## F = -230.441
## final value -230.440696
## converged
```

```
opt_true_A <- c(opt_mean[3]/60,opt_mean[1])
opt_true_B <- c(opt_mean[4]/60,opt_mean[2])
```

Plot Optimal Decision Uncertainty (Figure 4 in Section 4)

```
aa=rbind(cbind(x_gp[,2], x_gp[,4]/60), cbind(x_gp[,1], x_gp[,3]/60))
df_plot <- data.frame(Time=aa[,2], Temperature=aa[,1])

custom_breaks <- c(0, 0.00005, 0.0001, 0.00015, 0.0002, 0.0005, 0.001, 0.005, 0.01, 0.05, 0.1)
ggplot(df_plot, aes(x = Time, y = Temperature)) +
  geom_density_2d_filled(aes(fill = ..level..), breaks = custom_breaks) +
  scale_fill_brewer(palette = "PuBu", name = "Density Level") +
  geom_point(aes(x = opt_true_A[1], y = opt_true_A[2]), color = "red", size = 2.5) +
  geom_point(aes(x = opt_true_B[1], y = opt_true_B[2]), color = "red", size = 2.5) +
  geom_point(aes(x = 7.5, y = 140, xend = opt_true_A[1], yend = opt_true_A[2]), color = "red") +
  geom_point(aes(x = opt_true_A[1], y = opt_true_A[2], xend = opt_true_B[1], yend = opt_true_B[2]), color = "red") +
  geom_point(aes(x = opt_true_B[1], y = opt_true_B[2], xend = 200.385, yend = 140), color = "red") +
  ylim(140, 185) +
  xlim(0, 205) +
  theme_minimal() +
  theme(
    axis.title.x = element_text(size = 14),
    axis.title.y = element_text(size = 14),
    axis.text.x = element_text(size = 12),
    axis.text.y = element_text(size = 12),
    plot.margin = unit(c(0.5, 3, 0.5, 3), "cm")
  ) +
  labs(x = "Time", y = "Temperature")
```



Compute Sensitivity Scores

```
X1 <- x_gp[1:(M/2), ]
X2 <- x_gp[(M/2+1):M, ]
sobol_result <- sobolSalt(model = surrogate_y, X1, X2, scheme="A", nboot = 100)

ss_effects_gp <- c(sobol_results$original, sobol_results$original)
ss_effects_gp[ss_effects_gp < 0] <- 0
ss_effects_gp[ss_effects_gp > 1] <- 1
```

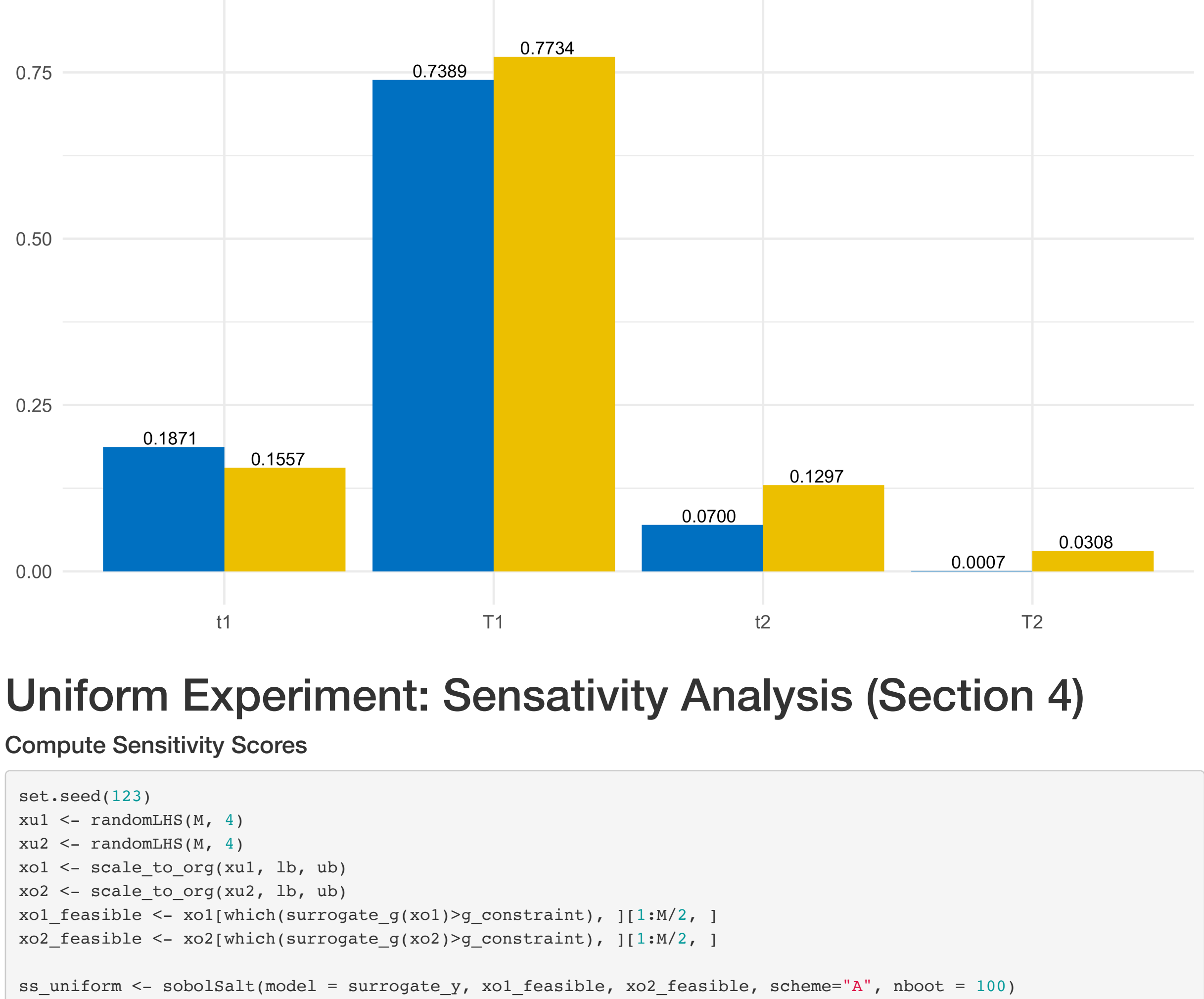
Plot Sensitivity Analysis Result From Decision Uncertainty Experiment (Figure 5 in Section 4)

```
sens_gp <- data.frame(X=rep(colnames(x.sample), 2), value = ss_effects_gp,
  group = rep(c('Sobol' First-Order Indices',
    'Sobol' Total Sensitivity Indices'),
    each=length(ss_effects_gp)/2))
```

```
p_gp_opt <- ggplot(sens_gp, aes(x = X, y = value)) +
  geom_col(aes(fill = group), position = "dodge") +
  geom_text(aes(label = sprintf("%.4f", value)), position = position_dodge2(width = 0.8),
  vjust = -0.2, size = 3)
```

```
geom_point(data = subset(sens_gp, group == "First"),
  position = position_dodge(width = 0.8), size = 4) +
  geom_point(data = subset(sens_gp, group == "Total"),
  position = position_dodge(width = 0.8), size = 4) +
  scale_color_manual(values = c("#0073C2FF", "#FFC000FF"), name = " ") +
  scale_y_continuous(limits = c(0, 1))
```

```
label_title = "Decision Uncertainty", x=NULL, y=NULL) +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5),
    legend.position = "top",
    axis.text = element_text(size = 9))
print(p_gp_opt)
```



## Uniform Experiment: Sensativity Analysis (Section 4)

Compute Sensitivity Scores

```
set.seed(123)
x1 <- randomJHS(M, 4)
x2 <- randomJHS(M, 4)
x0 <- scale_to_orig(x1, lb, ub)
x02 <- scale_to_orig(x2, lb, ub)
x01_feasible <- x01[which(surrogate_g(x01)>g_constraint), ][1:M/2, ]
x02_feasible <- x02[which(surrogate_g(x02)>g_constraint), ][1:M/2, ]

ss_effects_uniform <- sobolSalt(model = surrogate_y, x01_feasible, x02_feasible, scheme="A", nboot = 100)
ss_effects_uniform <- c(ss_effects_uniform$original, ss_effects_uniform$original)
ss_effects_uniform[ss_effects_uniform < 0] <- 0
ss_effects_uniform[ss_effects_uniform > 1] <- 1
```

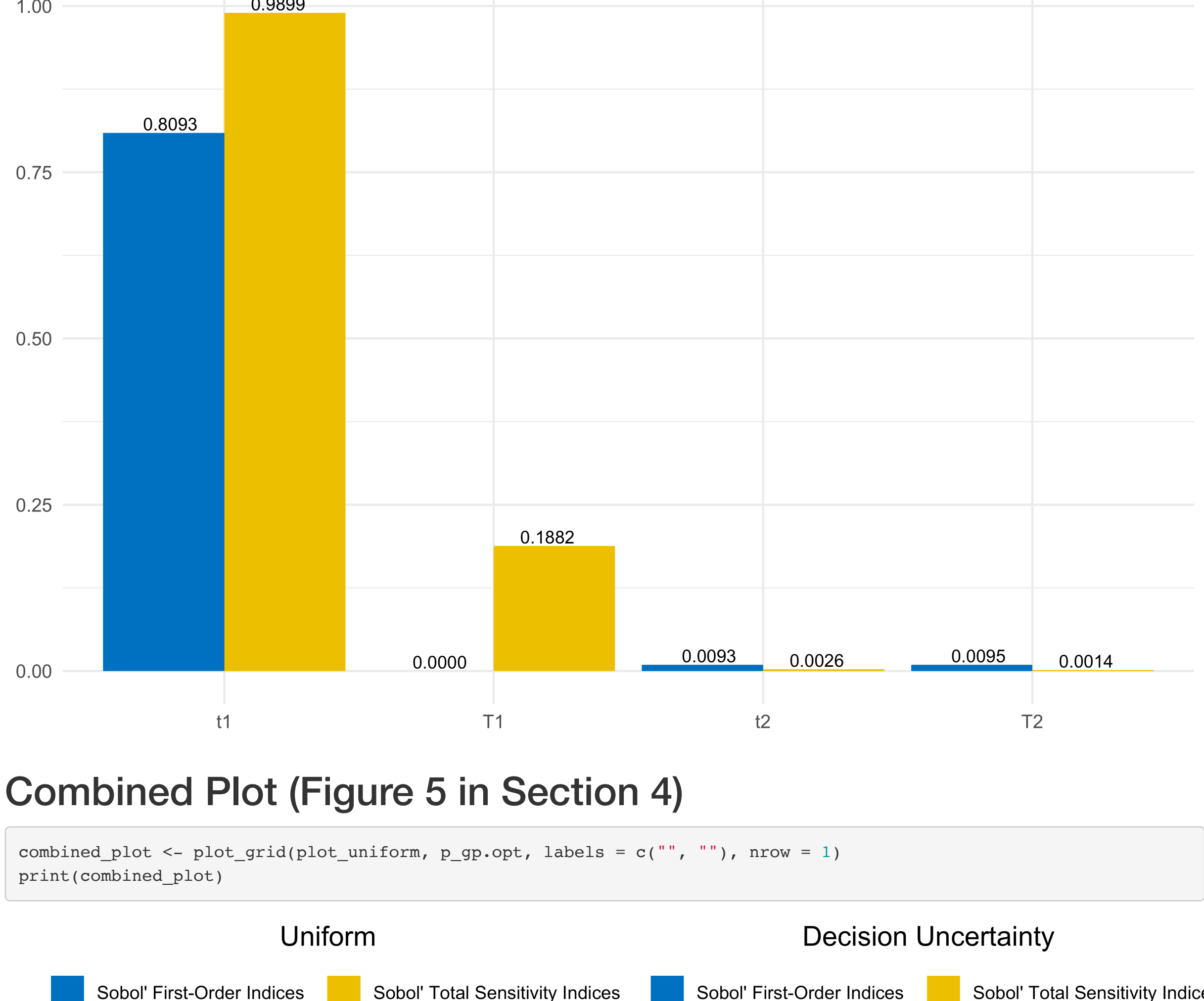
Plot Sensitivity Analysis Result From Uniform Experiment (Figure 5 in Section 4)

```
sens_uniform <- data.frame(X=rep(colnames(x.sample), 2), value = ss_effects_uniform,
  group = rep(c('Sobol' First-Order Indices',
    'Sobol' Total Sensitivity Indices'),
    each=length(ss_effects_uniform)/2))
```

```
plot_uniform <- ggplot(sens_uniform, aes(x = X, y = value)) +
  geom_col(aes(fill = group), position = "dodge") +
  geom_text(aes(label = sprintf("%.4f", value)), position = position_dodge2(width = 0.8),
  vjust = -0.2, size = 3)
```

```
geom_point(data = subset(sens_uniform, group == "First"),
  position = position_dodge(width = 0.8), size = 4) +
  geom_point(data = subset(sens_uniform, group == "Total"),
  position = position_dodge(width = 0.8), size = 4) +
  scale_color_manual(values = c("#0073C2FF", "#FFC000FF"), name = " ") +
  scale_y_continuous(limits = c(0, 1))
```

```
label_title = "Uniform", x=NULL, y=NULL) +
  theme_minimal() +
  theme(plot.title = element_text(hjust = 0.5),
    legend.position = "top",
    axis.text = element_text(size = 9))
print(plot_uniform)
```



## Combined Plot (Figure 5 in Section 4)

```
combined_plot <- plot_grid(plot_uniform, p_gp_opt, labels = c("", ""), nrow = 1)
print(combined_plot)
```

