

Modèle Quantal Response Equilibrium

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2025-05-24

Introduction

Ce document présente une estimation individuelle du modèle **Quantal Response Equilibrium (QRE)** basé sur les données collectées. Nous estimons pour chaque participant le paramètre λ , mesurant la sensibilité aux gains. L'ajustement du modèle est évalué par la log-vraisemblance et le score de Brier.

Chargement des bibliothèques et données

```
library(readxl)
library(ggplot2)
library(tidyr)
library(dplyr)
library(xtable)
library(factoextra)
```

```
donnees <- read_excel('../data/Exp_rience.xlsx')
donneesQ = donnees[1:63,7:21]
```

```
questions <- c("Q00_16-20 Arad & Rubinstein", "Q00_Mode du groupe", "Q00_ChatGPT",
               "Q00_Prime de 20", "Q01_12-20", "Q02_Alaoui & Penta", "Q03_Alaoui & Penta",
               "Q04_Goeree moderate", "Q05_Goeree extreme", "Q06_Cycle A", "Q07_Cycle C",
               "Q08_Pan A", "Q09_Bear C", "Q10_long pan A", "Q11_long pan C")
```

```
actions_par_question <- list(
  "Q00_16-20 Arad & Rubinstein" = 16:20,
  "Q00_Mode du groupe" = 16:20,
  "Q00_ChatGPT" = 16:20,
  "Q00_Prime de 20" = 16:20,
  "Q01_12-20" = c(12,14,16,18,20),
  "Q02_Alaoui & Penta" = c(12,14,16,18,20),
  "Q03_Alaoui & Penta plus" = c(12,14,16,18,20),
  "Q04_Goeree moderate" = c(14,12,18,16,20),
  "Q05_Goeree extreme" = c(18,16,14,12,20),
  "Q06_Cycle A" = c(12, 14, 16, 18, 20),
  "Q07_Cycle C" = c(12, 14, 16, 18, 20),
  "Q08_Pan A" = c(12, 14, 16, 18, 20),
  "Q09_Bear C" = c(12, 14, 16, 18, 20),
  "Q10_long pan A" = c(12, 14, 16, 18, 20),
  "Q11_long pan C" = c(12, 14, 16, 18, 20)
)
```

```
bonus_par_question <- list(
```

```

"Q00_16-20 Arad & Rubinstein" = list(gauche = 10, egal = 0),
"Q00_Mode du groupe" = list(gauche = 10, egal = 0),
"Q00_ChatGPT" = list(gauche = 10, egal = 0),
"Q00_Prime de 20" = list(gauche = 20, egal = 0),
"Q01_12-20" = list(gauche = 20, egal = 0),
"Q02_Alaoui & Penta" = list(gauche = 20, egal = 10),
"Q03_Alaoui & Penta plus" = list(gauche = 40, egal = 10),
"Q04_Goeree moderate" = list(gauche = 20, egal = 0),
"Q05_Goeree extreme" = list(gauche = 20, egal = 0),
"Q06_Cycle A" = list(gauche = 20, egal = 0),
"Q07_Cycle C" = list(gauche = 20, egal = 0),
"Q08_Pan A" = list(gauche = 20, egal = 0),
"Q09_Bear C" = list(gauche = 20, egal = 0),
"Q10_long pan A" = list(gauche = 20, egal = 0),
"Q11_long pan C" = list(gauche = 20, egal = 0)
)

```

Fonctions

```

logit_qre_response <- function(payoffs, lambda) {
  exp_payoff <- exp(lambda * payoffs)
  exp_payoff / sum(exp_payoff)
}

qre_strategy <- function(lambda, actions, bonus_gauche, bonus_egal) {
  expected_payoff <- sapply(seq_along(actions), function(a_idx) {
    a <- actions[a_idx]
    if (a_idx < length(actions)) {
      bonus_pos <- a_idx + 1
      bonus_left <- bonus_gauche
    } else {
      bonus_left <- 0
    }
    bonus_same <- bonus_egal
    gain <- a + bonus_left + bonus_same
    return(gain)
  })

  probs <- logit_qre_response(expected_payoff, lambda)
  return(probs)
}

```

Log-vraisemblance et estimation

```
log_likelihood_qre <- function(lambda, reponses, questions, actions_map) {
  total_loglik <- 0

  for (j in seq_along(reponses)) {
    r <- reponses[j]
    q <- questions[j]
    actions <- actions_map[[q]]
    bonus <- bonus_par_question[[q]]

    qre_probs <- qre_strategy(lambda, actions, bonus_gauche = bonus$gauche, bonus_egal =

    if (!is.na(r)) {
      idx <- which(actions == r)
      if (length(idx) == 1) {
        p_r <- qre_probs[idx]
      } else {
        p_r <- 0
      }
      if (p_r > 0 && !is.na(p_r)) {
        total_loglik <- total_loglik + log(p_r)
      } else {
        total_loglik <- total_loglik - 1e6
      }
    }
  }

  return(total_loglik)
}
```

Estimation individuelle

```
resultats_qre <- data.frame()

for (i in 1:(nrow(donnees)-1)) {
  individu <- donnees[i, ]

  reponses <- sapply(questions, function(q) {
    rep <- individu[[q]]
    if (!is.na(rep) && grepl(":", rep)) {
      val <- as.numeric(trimws(strsplit(rep, ":")[[1]][2]))
    } else {
```

```

    val <- as.numeric(trimws(rep))
  }
  return(val)
})

tryCatch({
  opt <- optim(
    par = c(lambda = 1),
    fn = function(par) -log_likelihood_qre(par[1], reponses, questions, actions_par_q
    method = "L-BFGS-B",
    lower = c(0.01), upper = c(20)
  )

  lambda_estime <- round(opt$par[1], 3)
  max_vraisemblance <- round(-opt$value, 3)

  # Score de Brier
  brier_total <- 0
  nb_valides <- 0

  for (j in seq_along(reponses)) {
    r <- reponses[j]
    q <- questions[j]
    actions <- actions_par_question[[q]]
    bonus <- bonus_par_question[[q]]

    qre_probs <- qre_strategy(lambda_estime, actions, bonus_gauche = bonus$gauche, bor

    if (!is.na(r)) {
      y_true <- rep(0, length(actions))
      idx <- which(actions == r)
      if (length(idx) == 1) {
        y_true[idx] <- 1
        brier_score <- sum((qre_probs - y_true)^2)
        brier_total <- brier_total + brier_score
        nb_valides <- nb_valides + 1
      }
    }
  }

  brier_moyen <- ifelse(nb_valides > 0, brier_total / nb_valides, NA)

  resultats_qre <- rbind(resultats_qre, data.frame(
    id = i,

```

```

        lambda = lambda_estime,
        VraisemblanceMax = max_vraisemblance,
        BrierScore = round(brier_moyen, 4)
    ))

}, error = function(e) {
    resultats_qre <- rbind(resultats_qre, data.frame(
        id = i,
        lambda = NA,
        VraisemblanceMax = NA,
        BrierScore = NA
    ))
    cat("Erreur pour l'individu", i, ":", e$message, "\n")
})
}
print(resultats_qre)

```

```

##           id lambda VraisemblanceMax BrierScore
## lambda    1  1.495          -4.312      0.1501
## lambda1    2  0.359         -16.690      0.6370
## lambda2    3  0.036         -23.806      0.7981
## lambda3    4  0.231         -20.010      0.7500
## lambda4    5  0.196         -20.648      0.7524
## lambda5    6  0.010         -22.613      0.8026
## lambda6    7  0.039         -23.769      0.7950
## lambda7    8  0.801         -11.084      0.4595
## lambda8    9  0.274         -19.254      0.7111
## lambda9   10  0.054         -23.492      0.7801
## lambda10  11  0.010         -24.914      0.8198
## lambda11  12  0.207         -20.447      0.7541
## lambda12  13  0.010         -24.194      0.8018
## lambda13  14  0.185         -20.838      0.6874
## lambda14  15  0.010         -24.414      0.8066
## lambda15  16  0.099         -19.577      0.7834
## lambda16  17  0.603         -13.861      0.5900
## lambda17  18  1.046           -8.347      0.3291
## lambda18  19  0.463         -14.107      0.5679
## lambda19  20  0.036         -23.806      0.7931
## lambda20  21  0.167         -21.190      0.6985
## lambda21  22  0.010         -24.924      0.8202
## lambda22  23  0.022         -24.007      0.7956
## lambda23  24  0.010         -25.064      0.8231
## lambda24  25  0.136         -21.794      0.7727
## lambda25  26  0.118         -22.175      0.7337

```

## lambda26	27	0.327	-18.353	0.6994
## lambda27	28	0.046	-20.486	0.7968
## lambda28	29	0.231	-20.010	0.7312
## lambda29	30	0.185	-20.838	0.6746
## lambda30	31	0.010	-24.214	0.8010
## lambda31	32	0.010	-25.974	0.8464
## lambda32	33	0.010	-25.374	0.8310
## lambda33	34	0.010	-25.974	0.8464
## lambda34	35	0.011	-24.104	0.7990
## lambda35	36	0.024	-23.984	0.7995
## lambda36	37	0.089	-22.790	0.7698
## lambda37	38	0.046	-23.642	0.7869
## lambda38	39	0.185	-20.838	0.7651
## lambda39	40	0.010	-24.424	0.8068
## lambda40	41	0.010	-24.154	0.8010
## lambda41	42	0.010	-24.264	0.8033
## lambda42	43	0.010	-24.494	0.8092
## lambda43	44	0.077	-23.038	0.7880
## lambda44	45	0.041	-23.729	0.7895
## lambda45	46	0.089	-22.790	0.7196
## lambda46	47	0.010	-24.134	0.7999
## lambda47	48	0.039	-23.769	0.7977
## lambda48	49	0.159	-21.353	0.7072
## lambda49	50	0.070	-23.184	0.7805
## lambda50	51	0.010	-24.184	0.8003
## lambda51	52	0.010	-24.874	0.8190
## lambda52	53	0.143	-21.655	0.7541
## lambda53	54	0.167	-21.190	0.7632
## lambda54	55	0.143	-21.655	0.7724
## lambda55	56	0.151	-21.508	0.6589
## lambda56	57	0.010	-24.124	0.7996
## lambda57	58	0.603	-13.861	0.5841
## lambda58	59	0.196	-20.648	0.7657
## lambda59	60	0.077	-23.038	0.7612
## lambda60	61	0.136	-21.794	0.6836
## lambda61	62	0.010	-25.884	0.8440
## lambda62	63	1.575	-4.552	0.1435

Visualisation

```
resultats_long <- resultats_qre %>%
  mutate(VraisemblanceMax_scaled = -VraisemblanceMax / 50) %>%
  pivot_longer(cols = c(lambda, VraisemblanceMax_scaled, BrierScore),
```

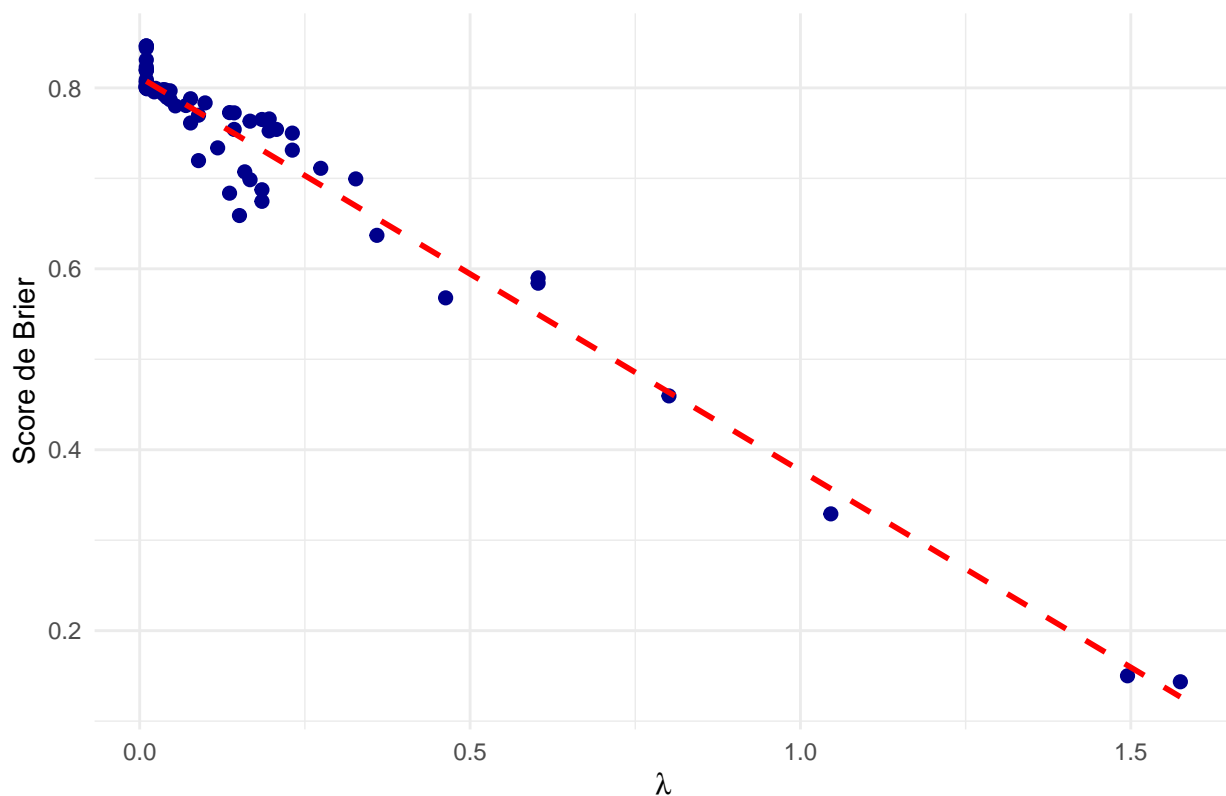
```

      names_to = "parametre", values_to = "valeur") %>%
mutate(parametre = recode(parametre,
                          "lambda" = "Lambda ( )",
                          "VraisemblanceMax_scaled" = "-Log-vraisemblance",
                          "BrierScore" = "Score de Brier"))

ggplot(resultats_qre, aes(x = lambda, y = BrierScore)) +
  geom_point(color = "darkblue", size = 2) +
  geom_smooth(method = "lm", se = FALSE, color = "red", linetype = "dashed") +
  theme_minimal() +
  labs(
    title = "",
    x = expression(lambda),
    y = "Score de Brier"
  )

```

```
## `geom_smooth()` using formula = 'y ~ x'
```



Estimation par question

```
resultats_par_question_qre <- data.frame()
```



```

for (q in questions) {
  actions <- actions_par_question[[q]]
  bonus <- bonus_par_question[[q]]

  reponses <- sapply(donnees[[q]], function(rep) {
    if (!is.na(rep) && grepl(":", rep)) {
      as.numeric(trimws(strsplit(rep, ":")[[1]][2]))
    } else {
      as.numeric(trimws(rep))
    }
  })

  tryCatch({
    opt <- optim(
      par = c(lambda = 1),
      fn = function(par) {
        lambda <- par[1]
        total_loglik <- 0
        for (j in seq_along(reponses)) {
          r <- reponses[j]
          if (is.na(r) || !(r %in% actions)) next
          probs <- qre_strategy(lambda, actions, bonus$gauche, bonus$egal)
          idx <- which(actions == r)
          p_r <- probs[idx]
          total_loglik <- total_loglik + ifelse(p_r > 0, log(p_r), -1e6)
        }
        return(-total_loglik)
      },
      method = "L-BFGS-B", lower = 0.01, upper = 20
    )

    lambda_estime <- round(opt$par[1], 3)
    loglik <- round(-opt$value, 3)

    # Score de Brier
    brier_total <- 0
    n_valid <- 0

    for (j in seq_along(reponses)) {
      r <- reponses[j]
      if (is.na(r) || !(r %in% actions)) next
      probs <- qre_strategy(lambda_estime, actions, bonus$gauche, bonus$egal)
      y_true <- rep(0, length(actions))
      idx <- which(actions == r)

```

```

    y_true[idx] <- 1
    brier_total <- brier_total + sum((probs - y_true)^2)
    n_valid <- n_valid + 1
  }

  brier_moyen <- if (n_valid > 0) round(brier_total / n_valid, 4) else NA

  resultats_par_question_qre <- rbind(resultats_par_question_qre, data.frame(
    Question = q,
    lambda = lambda_estime,
    LogVraisemblance = loglik,
    Brier = brier_moyen
  ))

}, error = function(e) {
  resultats_par_question_qre <- rbind(resultats_par_question_qre, data.frame(
    Question = q,
    lambda = NA,
    LogVraisemblance = NA,
    Brier = NA
  ))
  cat("Erreur pour la question", q, ":", e$message, "\n")
})
}

# Affichage final
print(resultats_par_question_qre)

```

```

##           Question lambda LogVraisemblance  Brier
## lambda  Q00_16-20 Arad & Rubinstein  0.102      -97.316 0.7814
## lambda1      Q00_Mode du groupe  0.065      -100.282 0.7924
## lambda2      Q00_ChatGPT  0.080      -99.778 0.7883
## lambda3      Q00_Prime de 20  0.042      -99.382 0.7885
## lambda4      Q01_12-20  0.032      -100.282 0.7924
## lambda5      Q02_Alaoui & Penta  0.082      -95.961 0.7577
## lambda6      Q03_Alaoui & Penta plus  0.016      -100.149 0.7928
## lambda7      Q04_Goeree moderate  0.026      -100.633 0.7944
## lambda8      Q05_Goeree extreme  0.044      -99.442 0.7836
## lambda9      Q06_Cycle A  0.010      -102.038 0.8038
## lambda10     Q07_Cycle C  0.010      -101.478 0.8004
## lambda11     Q08_Pan A  0.028      -100.524 0.7938
## lambda12     Q09_Bear C  0.010      -101.398 0.7999
## lambda13     Q10_long pan A  0.034      -93.830 0.7903
## lambda14     Q11_long pan C  0.019      -94.565 0.7973

```

Génération des graphiques

```
plots <- list()

for (q in 1:15) {
  actions <- actions_par_question[[q]]
  bonus <- bonus_par_question[[q]]

  reponses <- sapply(donneesQ[[q]], function(rep) {
    if (!is.na(rep) && grepl(":", rep)) {
      as.numeric(trimws(strsplit(rep, ":")[[1]][2]))
    } else {
      as.numeric(trimws(rep))
    }
  })

  tryCatch({
    opt <- optim(
      par = c(lambda = 1),
      fn = function(par) {
        lambda <- par[1]
        total_loglik <- 0
        for (j in seq_along(reponses)) {
          r <- reponses[j]
          if (is.na(r) || !(r %in% actions)) next
          probs <- qre_strategy(lambda, actions, bonus$gauche, bonus$egal)
          idx <- which(actions == r)
          p_r <- probs[idx]
          total_loglik <- total_loglik + ifelse(p_r > 0, log(p_r), -1e6)
        }
        return(-total_loglik)
      },
      method = "L-BFGS-B", lower = 0.01, upper = 20
    )

    lambda_estime <- opt$par[1]
    strategie_qre <- qre_strategy(lambda_estime, actions, bonus$gauche, bonus$egal)

    df_pred <- data.frame(
      Action = factor(actions, levels = actions),
      Valeur = strategie_qre,
      Type = "Prévu"
    )
  })
}
```

```

Q <- donneesQ[[q]]
Qr <- table(Q) / length(Q)
df_obs <- data.frame(
  Action = factor(actions, levels = actions),
  Valeur = as.numeric(Qr),
  Type = "Observé"
)

df_combined <- rbind(df_pred, df_obs)

p <- ggplot(df_combined, aes(x = Action, y = Valeur, fill = Type)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(title = paste("Question", q+1),
       x = "Action possible", y = "Fréquence") +
  scale_fill_manual(values = c("Observé" = "black", "Prévu" = "lightgrey")) +
  theme_minimal(base_size = 10) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        legend.position = "none")

plots[[length(plots) + 1]] <- p

}, error = function(e) {
  cat("Erreur pour la question", q, ":", e$message, "\n")
})
}

# Suppression de l'ancien PDF
pdf_path <- "comparaison_QRE_5x3.pdf"
if (file.exists(pdf_path)) file.remove(pdf_path)

## [1] TRUE

# Export en PDF
library(ggpubr)
pdf(pdf_path, width = 20, height = 12)
ggarrange(plotlist = plots, ncol = 5, nrow = 3)
dev.off()

## pdf
## 2

```