

# Moduł 12

## Przypadek użycia 2

### Analiza reakcji biochemicznych

iBigWorld:  
Innovations for Big Data in a Real World

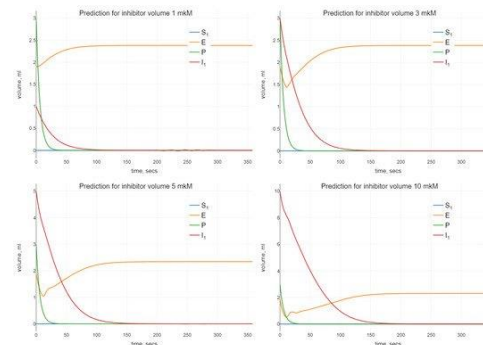
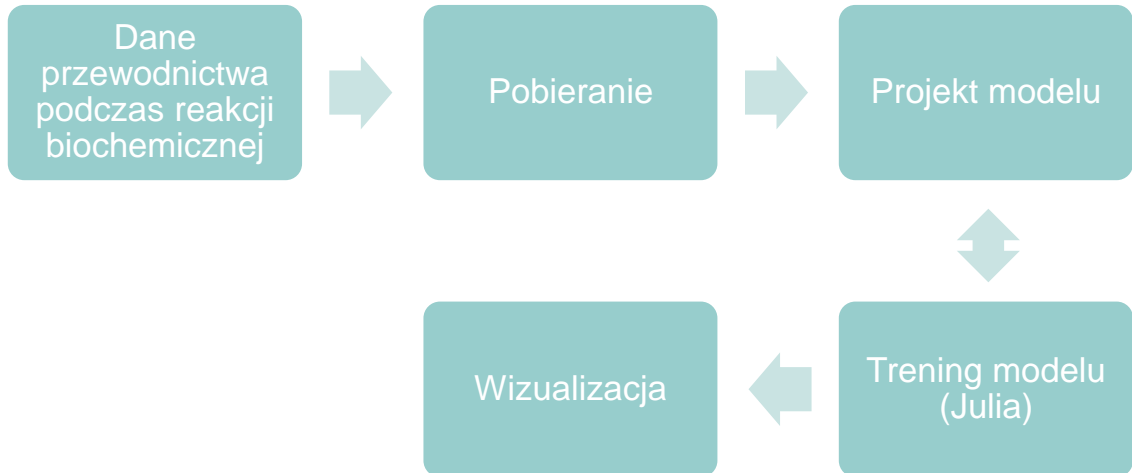
Zespół UBB

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# Identyfikacja parametrów reakcji biochemicznych

- Opracowano i przeanalizowano numerycznie i eksperymentalnie modele dynamiczne kompleksów enzym-substrat i enzym-substrat-inhibitor.
- Konstruujemy wielosubstratowy model wieloinhibitorowy.
- Opracowano algorytm identyfikacji parametrów, który został przetestowany na danych eksperymentalnych przewodnictwa roztworu i jest zgodny z Big Data.
- W Julii zaimplementowano obliczenia o wysokiej wydajności.



# Pobieranie danych

- *Interakcja enzym-substrat*
- Badania przeprowadzono w roztworach wodnych z użyciem następujących substancji: białko – BSA; enzym – enzym acetylocholinoesteraza (AChE); oraz substrat — chlorek acetylocholinyl (AChCl). Do kompleksu utworzonego z BSA 2 mg/ml (objętość 4 ml) i enzymem 2 mg/ml (objętość 0,1 ml) (kompleks jest znany jako usieciowany agregat enzymatyczny (CLEA)), substrat 2 mg /ml w różnych objętościach (0,1 ml, 0,3 ml, 0,9 ml, 1,5 ml, itd.). W efekcie otrzymano próbki:
  - Próbka 1. BSA + enzyme + 0.1 mL substratu;
  - Próbka 2. BSA + enzyme + 0.3 mL substratu;
  - Próbka 3. BSA + enzyme + 0.9 mL substratu;
  - Próbka 4. BSA + enzyme + 1.5 mL substratu;
  - itd...
- Przewodność kompleksu BSA + enzym + substrat badano za pomocą specjalnie skonstruowanego układu pomiarowego, w czasie 500 s z próbkowaniem sygnału co 5 s.



# Pobranie danych

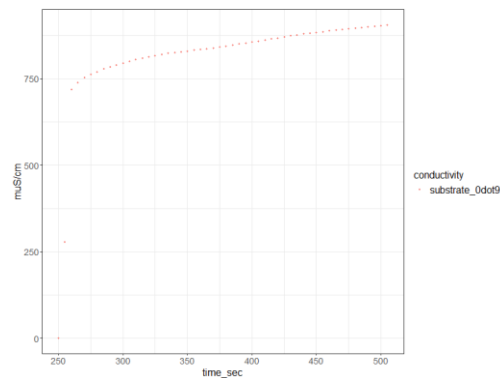
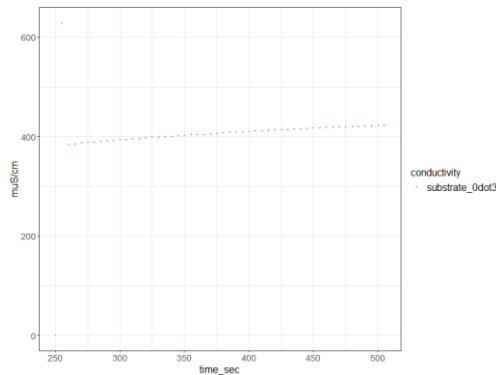
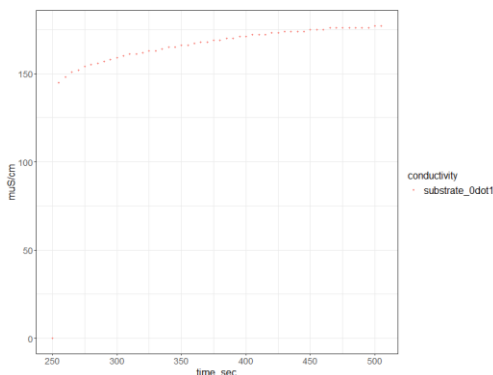
- `df<-read.xlsx(„conductivity.xls”,sheetName = "Sheet1",stringsAsFactors=F)`
- `df <- df[which(df$time_sec>=250),]`
- `tmpvec <- df$substrate_0dot1`
- `df$substrate_0dot1 <- df$substrate_0dot1 - c(rep(tmpvec[1],length(tmpvec)))`
- `tmpvec <- df$substrate_0dot3`
- `df$substrate_0dot3 <- df$substrate_0dot3 - c(rep(tmpvec[1],length(tmpvec)))`
- `tmpvec <- df$substrate_0dot9`
- `df$substrate_0dot9 <- df$substrate_0dot9 - c(rep(tmpvec[1],length(tmpvec)))`
- `tmpvec <- df$substrate1dot5`
- `df$substrate1dot5 <- df$substrate1dot5 - c(rep(tmpvec[1],length(tmpvec)))`

time_sec	substrate_0dot1	substrate_0dot3	substrate_0dot9	substrate1dot5
250	0	0.0	0.0	0.0
255	145	628.6	279.2	1291.4
260	148	383.6	719.2	1287.4
265	151	385.6	739.2	1301.4
270	152	387.6	753.2	1305.4
275	154	388.6	762.2	1308.4
280	155	389.6	770.2	1311.4
285	156	390.6	778.2	1314.4
290	157	391.6	784.2	1319.4
295	158	392.6	790.2	1324.4
300	159	393.6	795.2	1331.4
305	160	394.6	801.2	1337.4
310	161	395.6	805.2	1345.4
315	161	396.6	809.2	1351.4
320	162	397.6	813.2	1356.4
325	163	398.6	817.2	1362.4
330	163	399.6	820.2	1367.4
335	164	400.6	823.2	1373.4
340	165	400.6	825.2	1377.4
345	165	401.6	828.2	1381.4
350	166	402.6	830.2	1385.4
355	166	403.6	832.2	1388.4
360	167	404.6	834.2	1391.4



# Eksploracja danych

- `tmp=melt(df,id.vars=c("time_sec"),variable.name="conductivity",value.name="muS/cm")`
- `png(paste(paste("substrate_0dot1"),"png", sep = "."), width = 800, height = 600)`
- `p<-ggplot(data=tmp[which(tmp$conductivity=="substrate_0dot1"),],aes(x=time_sec,y=`muS/cm`,color=conductivity))+geom_point(size=1)+my_theme`
- `plot(p)`
- `dev.off()`
- `png(paste(paste("substrate_0dot3"),"png", sep = "."), width = 800, height = 600)`
- `p<-ggplot(data=tmp[which(tmp$conductivity=="substrate_0dot3"),],aes(x=time_sec,y=`muS/cm`,color=conductivity))+geom_point(size=1)+my_theme`
- `plot(p)`
- `dev.off()`
- `png(paste(paste("substrate_0dot9"),"png", sep = "."), width = 800, height = 600)`
- `p<-ggplot(data=tmp[which(tmp$conductivity=="substrate_0dot9"),],aes(x=time_sec,y=`muS/cm`,color=conductivity))+geom_point(size=1)+my_theme`
- `plot(p)`
- `dev.off()`
- `png(paste(paste("substrate_1dot5"),"png", sep = "."), width = 800, height = 600)`
- `p<-ggplot(data=tmp[which(tmp$conductivity=="substrate_1dot5"),],aes(x=time_sec,y=`muS/cm`,color=conductivity))+geom_point(size=1)+my_theme`
- `plot(p)`
- `dev.off()`



# Przetwarzanie wysokowydajne w Julia

```
• library(JuliaCall)
• library(diffeqr)
• de <- diffeqr::diffeq_setup()
• f <- JuliaCall::julia_eval("function f(du, u, h, p, t)
•     k_d=p[1]
•     a=p[2]
•     m=p[3]
•     tau_m=p[4]
•     n_S_0=p[5]
•     n_E_0=p[6]
•     n_P_0=p[7]
•     n=500
•     rmse=sqrt((m+1)/a^2)
•     lags=range(1/n, length=n, stop=(tau_m+(m+1)/a+3*rmse)*(1-1/n))
•     du[1]= -k_d*u[2]*u[1]
•     du[2]=-k_d*u[2]*u[1] + k_d*(tau_m+(m+1)/a+3*rmse)*(1/n)*sum([psi(a,m,tau_m,tau)*h(p,t-tau)[2]*h(p,t-tau)[1] for tau in lags])
•     du[3]=k_d*(tau_m+(m+1)/a+3*rmse)*(1/n)*sum([psi(a,m,tau_m,tau)*h(p,t-tau)[2]*h(p,t-tau)[1] for tau in lags])
• end")
• h <- JuliaCall::julia_eval("function h(p, t)
•     [p[5],p[6],p[7]]
• end")
• psi <- JuliaCall::julia_eval("function psi(a, m, tau_m, tau)
•     if tau <= tau_m
•         res = 0
•     else
•         res = (a^(m+1)/gamma(m+1))*(tau-tau_m)^m * exp(-a*(tau-tau_m))
•     end
•     res
• end")
```



# Przetwarzanie wysokowydajne w Julia (kontynuacja)

- `u0 <- c(0.3,1,0)`
- `tspan <- c(max(0,min(df$time_sec)),max(0,max(df$time_sec)))`
- `abstol <- 1e-2`
- `reltol <- 1e-2`
- `saveat <- round(seq(from=min(df$time),to=max(df$time),length.out=500))`
- `saveat <- unique(saveat)`
- `p <- c(0.05,1,20,5,u0[1:3])`
- `constant_lags <- c(100.0)`
- `JuliaCall::julia_assign("u0", u0)`
- `JuliaCall::julia_assign("tspan", tspan)`
- `JuliaCall::julia_assign("constant_lags", constant_lags)`
- `JuliaCall::julia_assign("p", p)`
- `prob <- JuliaCall::julia_eval("using SpecialFunctions; DDEProblem(f, u0, h, tspan, p)")`
- `system.time(sol <- de$solve(prob,abstol=abstol,reltol=reltol, saveat=saveat, de$MethodOfSteps(de$Tsit5())))`
- `udf <- as.data.frame(cbind(sol$t,t(sapply(sol$u,identity))))`
- `colnames(udf) <- c("t","S","E","P")`

# Strojenie modelu

- Implementacja algorytmu COBYLA do problemu

**Input data:**  $X_{exp}, \Pi_{lower}, \Pi_{upper}, \Pi_{init}$

**Result:**  $\Pi_{opt}$

- 1 form the initial simplex  $C_{init}$  with the vertices  $\Pi_0^{init}, \Pi_1^{init}, \dots, \Pi_6^{init}$ ;
- 2 **repeat**
- 3     for the current simplex  $C$  calculate the values  $\hat{\Phi}_C(\Pi_i), i = \overline{0, 6}$ ;
- 4     search the vertex  $\Pi_p$  determined by the equation  

$$\hat{\Phi}_C(\Pi_p) = \min\{\hat{\Phi}_C(\Pi_i), i = \overline{0, 6}\};$$
- 5     calculate new vertex as  $\Pi_{new} := -\theta\Pi_p + (1 + \theta)\frac{1}{6}\sum_{i=\overline{0,6}, i \neq p} \Pi_i$ , where reflection coefficient  $\theta \in (0, 1)$  being chosen as small as possible in order  $\hat{\Phi}_C(\Pi_p)$  not were the least calculated function value so far;
- 6     form modified simplex  $C_{new}$  replacing vertex  $\Pi_p$  with  $\Pi_{new}$ ;
- 7     search  $\Pi_{opt}$  as a solution of the problem of linear optimization

$$\left. \begin{array}{l} \text{minimize} \quad \hat{\Phi}_{C_{new}}(\Pi), \\ \text{subject to} \quad \Pi \in C_{new} \end{array} \right\} \quad (21)$$

8 **until** stop condition;

9 **return**  $\Pi_{opt}$ ;





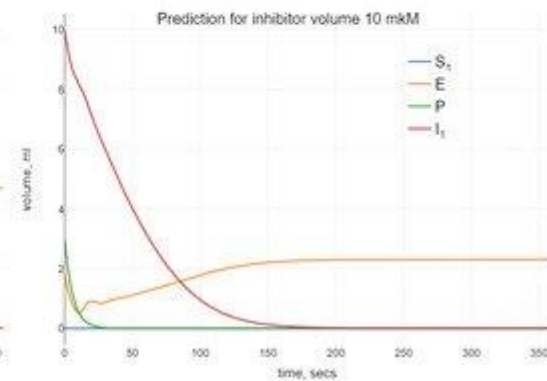
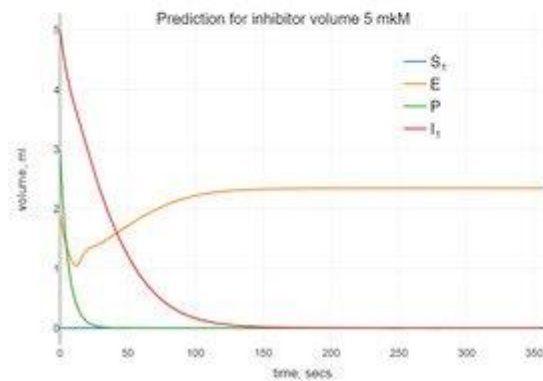
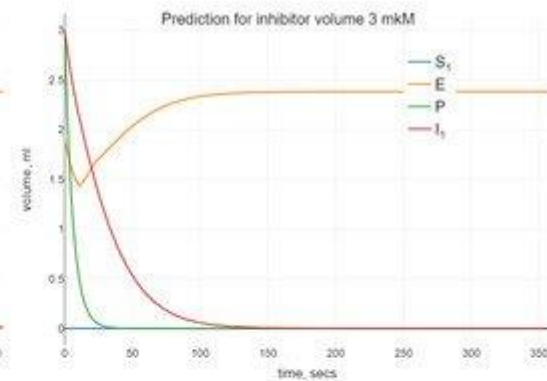
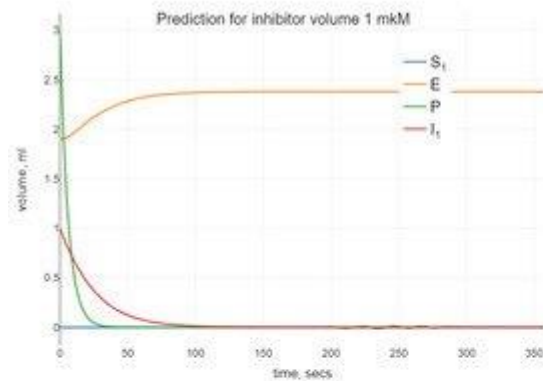
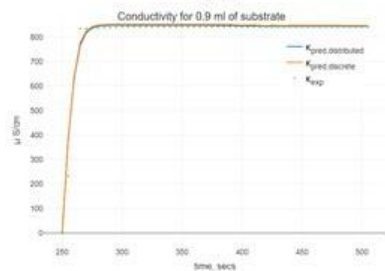
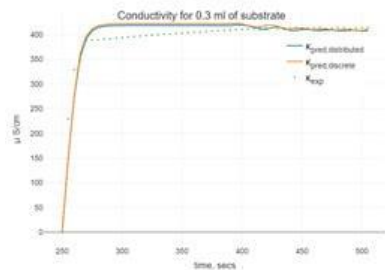
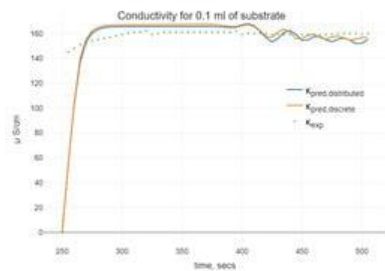
# Strojenie modelu

- Implementacja algorytmu COBYLA do problemu

- # M. J. D. Powell, "A direct search optimization method that models the objective and constraint functions by linear interpolation," in Advances in Optimization # and Numerical Analysis, eds. S. Gomez and J.-P. Hennart
- library(nloptr)
- fn.sir <- function(x) {
- sqrt(sum(ssq(x)^2))
- }
- hin.sir <- function(x) {
- h <- numeric(61)
- h[1:29] <- upper - x
- h[30:59] <- x - lower
- h[60] <- x[14]-x[15]-x[16]-x[17]-x[18]-x[19]-x[20]-x[21]
- h[61] <- -x[14]+x[15]+x[16]+x[17]+x[18]+x[19]+x[20]+x[21]
- #h[2] <- -x[14]+x[15]+x[16]+x[17]+x[18]+x[19]+x[20]+x[21]
- return(h)
- }
- fitval.cobyala <- cobyala(parms, fn.sir, hin = hin.sir,
- nl.info = TRUE, control = list(xtol\_rel = 1e-3,maxeval = 100))
- 
- parest <- fitval.cobyala\$par
- 
- fitval.cobyala\$value



# Wizualizacja



# Wizualizacja

```

library(plotly)
fig <- plotly::plot_ly(udf, x = sol$t, y = ~S, name="S", type =
'scatter', mode = 'lines') %>% plotly::add_trace(y = ~E,
name="E") %>% plotly::add_trace(y = ~P,name="P") %>%
  layout(
    title = list(text="Prediction for substrate volume 0.1
ml",y=0.95),
    xaxis = list(title = "time, secs"),
    yaxis = list(title = "volume, ml"),
    font = font
  )
orca(fig, file = paste(paste("Prediction_0dot1"),"png", sep =
"."))

```

