# R-code for producing Figure 2 from Pedersen et al. (2013)

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This vignette includes the commands to reproduce Fig. 2 from "A kinetic model explains why shorter and less affine enzyme-recruiting oligonucleotides can be more potent". The R-functions from the ASOmodels package are used.

```
> require(devtools)
> #install_github(ASOmodel,username=lykkep,build=FALSE)
> require(ASOmodels)
```

### Kinetic model figures

# Figure 2a: Time-resolved simulation of the model (unsaturated)

Parameters for the model, the initial concentrations and the time-steps for which the simulation is performed:

Using vode() the model is simulated in time. The function diffASO() is part of the ASOmodels package.

```
> solASO <- vode(init,0:100,diffASO,parms)</pre>
```

The timetraces for the concentrations of [O], [T], [OT], [OTE], and [E] are plotted:

```
> colVAR <- c("black", "darkgreen", "darkred", "orange", "green")
> SSvalue <- signif(last(solASO),3)
> SSvalue[c(3:4,6)] <- SSvalue[c(3:4,6)]*1E3
> xtime <- 59; ySS <- c(0.08,0.88,0.83,0.18,0.13)
> uSS <- c(nM,pM,pM,nM,pM)
> labSS=c(T,OT,OTE,E,O)
```

```
> par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
      las=1,cex.lab=1.25)
> plot(0,0,ylim=c(0,1),xlim=c(0,xtime+30),type=n,xaxt=n,
       yaxt=n,ylab=relative concentrations,xlab=minutes)
> axis(2,at=c(0,1),label=c(min,max),las=1)
> axis(1,at=c(seq(0,xtime,by=10),xtime+20),
       label=c(seq(0,xtime,by=10),))
> axis(1,at=xtime+20,label=steady-\nstate,mgp=c(0,1.6,0))
> for(i in 2:6){
    par(new=TRUE)
    plot(solASO[1:(xtime+1),1], solASO[1:(xtime+1),i], xaxt=n,
         yaxt=n,ylim=range(solASO[,i]),col=colVAR[i-1],
         type=1,ylab=NA,xlab=NA,xlim=c(0,xtime+30))
    lines(xtime+5:37,rep(solASO[101,i],33),col=colVAR[(i-1)])
    y <- diff(par(usr)[3:4])*ySS[i-1]+par(usr)[3]</pre>
    text(xtime+5, y, col=colVAR[i-1], adj=0, cex=0.8,
         substitute(b==e*1,list(e=SSvalue[i],l=uSS[i-1],b=labSS[i-1]) ))
+ }
```

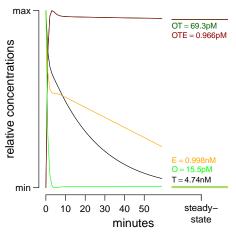


Figure 2a: Time resolved simulation of the relative concentrations of key species.

# Figure 2b: Time-resolved simulation of the model (saturated)

The initial oligonucleotide concentration is increased to 100nM.

The timetraces for the concentrations of [O], [T], [OT], [OTE], and [E] are plotted:

```
> SSvalue <- signif(last(solASO),3)</pre>
> SSvalue[c(2,4)] \leftarrow SSvalue[c(2,4)]*1E3
> xtime <- 55; ySS <- c(0.06,0.3,0.34,0.26,0.67)
> uSS \leftarrow c(pM,nM,pM,rep(nM,2))
> labSS=c(T,OT,OTE,E,O)
> par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
      las=1, cex.lab=1.25)
> plot(0,0,ylim=c(0,1),xlim=c(0,xtime+25),type=n,xaxt=n,
       yaxt=n,ylab=relative concentrations,xlab=minutes)
> axis(2,at=c(0,1),label=c(min,max),las=1)
> axis(1,at=c(seq(0,xtime,by=10),xtime+15),
       label=c(seq(0,xtime,by=10),))
> axis(1,at=xtime+15,label=steady-\nstate,mgp=c(0,1.6,0))
> for(i in 2:6){
    par(new=TRUE)
    plot(solASO[1:(xtime+1),1], solASO[1:(xtime+1),i], xaxt=n,
         yaxt=n,ylim=range(solASO[,i]),col=colVAR[i-1],
         type=1,ylab=NA,xlab=NA,xlim=c(0,xtime+25))
    lines(xtime+5:25,rep(solASO[101,i],21),col=colVAR[(i-1)])
    y <- diff(par(usr)[3:4])*ySS[i-1]+par(usr)[3]</pre>
    text(xtime+5,y,col=colVAR[i-1],adj=0,cex=0.8,
         substitute(b==e*1,list(e=SSvalue[i],l=uSS[i-1],b=labSS[i-1])))
+ }
```

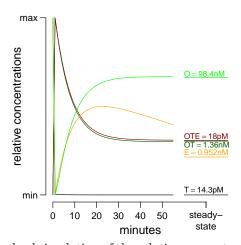


Figure 2b: Time resolved simulation of the relative concentrations of key species.

### Figure 2c: Simulated dose-response curve

Given a set of parameters the R-function Trel() from the ASOmodels package calculates the relative target concentration as a function of the total concentration of oligonucleotide added to the system.

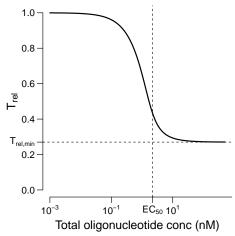


Figure 2c: The relative total target concentration  $(T_{\rm rel})$  is defined as the steady state level of total target in the presence of oligonucleotide divided by the target concentration in the absence of oligonucleotide. Dashed lines indicate 1-efficacy (horizontal) and  $EC_{50}$  (vertical).

#### Figure 2d: An optimal affinity

For a range of affinities  $D1\_seq$  the  $EC_{50}$ -values are calculated by use of the R-function EC50() from the ASOmodels package:

```
> D1_seq <- 10^seq(-3,3,by=0.25)
> ECfit <- sapply(D1_seq,EC50)</pre>
```

When there is no coupling between the off-rates  $k_{\text{OT}\to\text{O+T}}$  and  $k_{\text{OC}\to\text{O+C}}$  then the value of  $k_{\text{OC}\to\text{O+C}}$  is set in the param vector as the entry 'kC':

```
> parmsN0 <- c(parms,kC=parms[kOpT]*parms[KdOT]/parms[alpha])

> names(parmsN0)[length(parmsN0)] <- kC

> ECfitN0 <- sapply(D1\_seq,EC50N0) #EC50 without coupling

For the range of affinities the corresponding EC_{50}-values are plotted:

> par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),

+ las=1,cex.lab=1.25)

> plot(D1\_seq,ECfit,log=xy,yaxt=n,type=1,xaxt=n,

+ xlab=expression(K[dOT]^n(nM)),

+ ylab=expression(EC[50]^n(nM)))

> lines(D1\_seq,ECfitN0,lty=2)

> axis(2,at=c(2,20,200),labels=c(2,20,200),las=2)

> axis(1,at=10^pretty(log10(D1\_seq)),

+ labels=pretty10expLP(10^pretty(log10(D1\_seq)),drop.1=T),)

> legend(topleft,c(Coupling,No coupling),lty=c(1,2),bty=n)
```

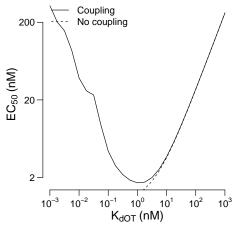


Figure 2d:  $EC_{50}$  as a function of the dissociation constant for the OT complex. A low  $K_{\rm dOT}$  corresponds to a high affinity binding. Dashed line: no coupling of off-rates. Solid line: coupling of off-rates.

# Experimental data figures

Figure 2e: Frieden et al. (2003)

```
> data(gapmers)
> dat <- data.frame(gapmers)
> colL <- c(red,orange,darkgreen,,darkblue,,</pre>
```

```
purple,,black)
> OLength <- sort(unique(dat$Oligo.length))</pre>
> #### We plot the data from Frieden et al, 2003
> dat.F <- dat[dat$Study=="Frieden 2003",]</pre>
> Flength <- dat.F$Oligo.length
  par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
      las=1, cex.lab=1.25)
> fitpar <- function(x,y){</pre>
    Fit \leftarrow lm(y x + I(x^2))
    coef <- coefficients(Fit)</pre>
    f <- summary(Fit)$fstatistic</pre>
    p <- signif(pf(f[1],f[2],f[3],lower.tail=F),2)</pre>
    xmin <- round(-coef[2]/(2*coef[3]))</pre>
    tmp <- as.expression(substitute(Optimal~T[m]%~~%x*degree*C,list(x=xmin)))</pre>
    return(list(coef=coef,p=p,legend=tmp))
+ }
> Fx <- dat.F$Predicted.Tm; Fy <- dat.F$Dose.2nm
> FitF <- fitpar(Fx,Fy)</pre>
> Parfun <- function(x){FitF$coef[1]+FitF$coef[2]*x+FitF$coef[3]*x^2}
  curve(Parfun(x),min(Fx),max(Fx), lwd=1,col=grey,ylim=c(0,110),
       ylab=activity (% of control),
       xlab=expression(predicted~T[m]~(*degree*C)))
  legend(diff(par(usr)[1:2])/2+par(usr)[1],100,
         c(Luciferase,paste(p =,FitF$p),FitF$legend),
         bty=n, cex=1.1, yjust=0.4, xjust=0.5)
> points(Fx,Fy,pch=19,cex=2,col=colL[Flength-11])
```

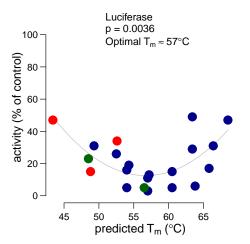


Figure 2e: 21 oligonucleotides targeted against the luciferase firefly gene.

# Figure 2f: Stanton et al. (2012)

```
> #### We plot the data from Stanton et al 2012
> dat.S <- dat[dat[,1]=="Stanton 2012",]</pre>
> Slength <- dat.S$Oligo.length
 par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
      las=1,cex.lab=1.25)
> Sx <- dat.S$Predicted.Tm; Sy <- dat.S$Dose.3nm
> FitS <- fitpar(Sx,Sy)</pre>
> Parfun <- function(x){FitS$coef[1]+FitS$coef[2]*x+FitS$coef[3]*x^2}
> curve(Parfun(x),min(Sx),max(Sx), lwd=1,col=grey,,ylim=c(0,110),
       ylab=mRNA (% of control),
       xlab=expression(predicted~T[m]~(*degree*C)))
> points(Sx,Sy, pch=19,col=colL[Slength-11],cex=2)
> legend(bottomright, as.character(sort(unique(OLength))),
         pch=19,col=colL[sort(unique(OLength))-11],bg=white,
         horiz=T, bty=n, cex=1.1
> legend(diff(par(usr)[1:2])/2+par(usr)[1],100,
         c(GR,paste(p =,FitS$p),FitS$legend),
         bty=n,cex=1.1,yjust=0.4,xjust=0.5 )
```

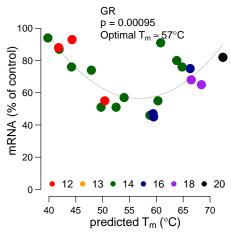


Figure 2f: 21 oligonucleotides targeted against the glucocorticoid receptor.

# Figure 2g: Pedersen et al. (2013) (this work)

```
> ### We plot the data from Pedersen et al, 2013
> dat.P <- dat[dat$Study=="Pedersen 2013",]
> Plength <- dat.P$Oligo.length
> par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
```

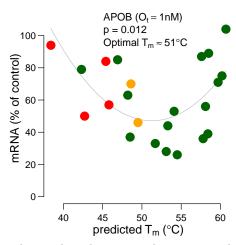


Figure 2g: 23 oligonucleotides targeted against apolipoprotein B.

# Figure 2h: Pedersen et al. (2013) (this work)

```
> ### We plot the data from Pedersen et al, 2013
> dat.P <- dat[dat$Study=="Pedersen 2013",]
> Plength <- dat.P$0ligo.length
> par(mar=c(3.2,3.4,0.1,0.1),bty=n,mgp=c(2,0.7,0),
+ las=1,cex.lab=1.25)
> Px <- dat.P$Predicted.Tm; Py <- dat.P$Dose.25nm
> FitP <- fitpar(Px,Py)
> Parfun <- function(x){FitP$coef[1]+FitP$coef[2]*x+FitP$coef[3]*x^2}
> curve(Parfun(x),min(Px),max(Px), lwd=1,col=grey,
+ xlab=expression(predicted~T[m]~(*degree*C)),
+ ylim=c(0,110),ylab=mRNA (% of control) )
> points(Px,Py,pch=19,cex=2,col=colL[Plength-11],)
> legend(diff(par(usr)[1:2])/2+par(usr)[1],100,
```

```
+ c(expression(APOB (*0[t]==25nM)),paste(p =,FitP$p),FitP$legend),
+ bty=n,cex=1.1,yjust=0.4,xjust=0.5 )
```

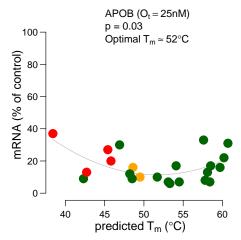


Figure 2h: 23 oligonucleotides targeted against apolipoprotein B.