

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

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With the gain of increasing reproducibility, this vignette includes the commands to reproduce Fig. 2 from “A kinetic model of enzyme recruiting oligonucleotides predicts an optimal affinity and thus explains why shorter and less affine oligonucleotides may be more potent”. The R-functions from the ASO-models are used and the package is loaded by the commands

```
> require(devtools)
> install_github('ASOmodel',username='lykkep')
> require(ASOmodels)
```

## Kinetic model figures

### Figure 2a: Time-resolved simulation of the ASO model

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

```
> parms <- c(Et = 1,KdOT = 0.3,kOpT = 0.2,KdOTE = 70,kOTpE = 5,
+           vprod = 0.2,vdegrad = 0.04,alpha=0.1,kcleav = 8)
> init <- c(T=parms['vprod']/parms['vdegrad'], OT=0, OTE=0,
+           E=parms['Et'], O=100, OCE=0, OC=0)
> TimeSteps <- c(seq(0,4.3,by=5E-2),seq(5,65,by=1))
```

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

```
> solASO <- vode(init,TimeSteps,diffASO,parms)
```

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

```
> SSvalue <- signif(last(solASO)[-1],2)
> solASO <- apply(solASO[,2:8],2,
+               function(x) (x-min(x))/max(x-min(x)) )
> colVAR <- c('black','darkgreen','darkred','orange','green')
```

```

> xtime <- TimeSteps <= 35
> par(mar=c(3.2,3.4,0.1,0.1),bty='n',mgp=c(2,0.7,0),
+     cex=0.6,cex.axis=1,las=1)
> for(i in 1:5){
+   if(i!=1) par(new=TRUE)
+   plot(TimeSteps[xtime], solASO[xtime,i], yaxt='n', xaxt='n',
+        ylab='relative concentrations', xlab='minutes',las=1,
+        col=colVAR[i], type='l', ylim=c(0,1), xlim=c(0,35+26))
+ }
> xtime <- 40
> for(i in 1:5) lines(xtime+0:20,rep(last(solASO)[i],21),
+                    col=colVAR[i])
> axis(1,at=c((0:3)*10,45),label=c((0:3)*10,''))
> axis(1,at=45,label='steady-\nstate',mgp=c(0,1.6,0))
> axis(2,at=c(0,1),label=c('min','max'),las=1)
> #O
> text(xtime,last(solASO)[5]-0.05,col=colVAR[5],adj=0,
+      substitute(O == e~nM,list(e=SSvalue[5])))
> #T
> text(xtime,0.05,col=colVAR[1],adj=0,
+      substitute(T == e*pM,list(e=1e3*SSvalue[1])))
> #OT
> text(xtime,last(solASO)[2]-0.05,col=colVAR[2],adj=0,
+      substitute(OT== e*nM,list(e=SSvalue[2])))
> #OTE
> text(xtime,last(solASO)[3]+0.05,col=colVAR[3],adj=0,
+      substitute(OTE == e*pM,list(e=1e3*SSvalue[3])))
> #E
> text(xtime,last(solASO)[4]+0.05,col=colVAR[4],adj=0,
+      substitute(E == e*nM,list(e=SSvalue[4])))

```

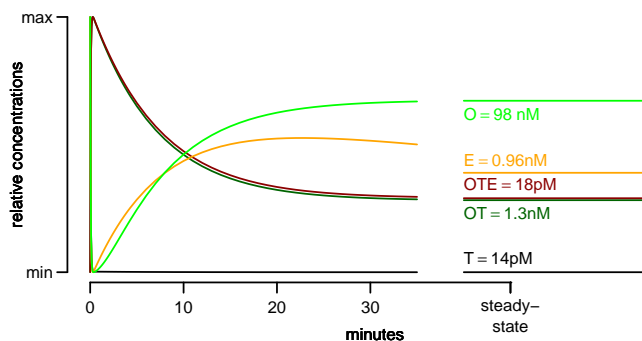


Figure 2a: Time resolved simulation of the relative concentrations of key species

## Figure 2b: 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.

```
> curve(Trel,1E-3,5E2,log='x', lwd=2,ylim=c(0,1),
+       ylab=expression(T[rel]),xaxt='n',
+       xlab='Total oligonucleotide conc (nM)')
> abline(h=Trel(1E9),lty=2) #Trel,min
> abline(v=IC50(parms['KdOT']),lty=2) #IC50
> axis(1,at=10^c(-3,-1,1,3),
+       labels=pretty10expLP(10^c(-3,-1,1,3),drop.1=T))
> axis(1,at=IC50(parms['KdOT']),label=expression(IC[50]))
> axis(2,at=Trel(1E6),
+       label=expression(T[rel*','*min]),las=1)
```

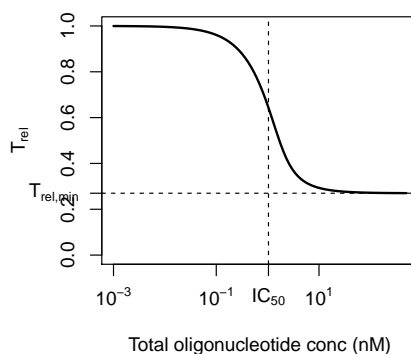


Figure 2b: The relative total target concentration ( $T_{rel}$ ) is defined as the steady state level of total target in the presence of oligonucleotide divided by the target concentration in the absence of oligonucleotide. Dashed lines indicate efficacy (horizontal) and  $IC_{50}$  (vertical).

## Figure 2c: An optimum affinity

For a sequence of affinities `D1_seq` the  $IC_{50}$  values are calculated by use of the R-function `IC50` from the ASOmodels package:

```
> D1_seq <- 10^seq(-3,3.2,by=0.25)
> ICfit <- sapply(D1_seq,IC50)
```

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

```

> parmsNO <- c(parms,kC=parms['kOpT']*parms['KdOT']/parms['alpha'])
> names(parmsNO)[length(parmsNO)] <- 'kC'
> ICfitNO <- sapply(D1_seq,IC50NO) #IC50 without coupling

```

For the sequence of affinities the sequences of  $IC_{50}$  values are plotted:

```

> plot(D1_seq,ICfit,log='xy',yaxt='n',type='l',xaxt='n',
+       xlab=expression(D[OT]~'(nM)'),ylab=expression(IC[50]~'(nM)'))
> lines(D1_seq,ICfitNO,lty=2)
> axis(2,at=c(2,20,200),labels=c(2,20,200))
> 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')

```

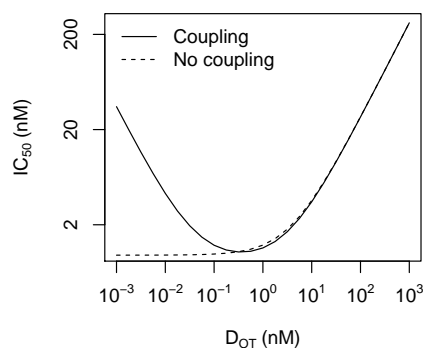


Figure 2c: The  $IC_{50}$  as a function of the dissociation constant for the OT complex. A low  $K_{dOT}$  corresponds to a high affinity binding. Dashed line: no coupling of off-rates. Solid line: coupling of off-rates.

## Experimental data figures

### Figure 2d: Frieden et al. (2003)

```

> data(gapmers)
> dat <- data.frame(gapmers)
> coll <- c('red','orange','darkgreen','','darkblue','','purple','','black')
> #### We plot the data from Frieden et al, 2003
> dat.F <- dat[dat$Study=="Frieden 2003",]
> cohig.F <- 63; colow.F <- 53
> tmp <- abs(cohig.F-colow.F)
> cut.F <- cut(dat.F$Predicted.Tm,c(0,colow.F,cohig.F,100),labels=F)

```

```

> Fx <- lapply(1:3,function(i)dat.F$Predicted.Tm[cut.F==i])
> Fy <- lapply(1:3,function(i) dat.F$Dose.2nm[cut.F==i])
> Flength <- dat.F$Oligo.length
> bp <- barplot(sapply(Fy,mean),ylim=c(0,55), las=1, axes=F,
+               yaxs='i',xaxs='i',space=0.01)
> plotCI(bp[,1],sapply(Fy,mean),sapply(Fy,sd),
+        add=T,pch=NA, gap=0,yaxs='i')
> par(new=T)
> plot(unlist(Fx),unlist(Fy),xlim=c(colow.F-tmp,cohigh.F+tmp),
+      ylim=c(0,55), pch=19, col=colL[Flength-11],xaxt='n',
+      ylab='% target measured from luciferase',
+      xlab=expression(T[m]~'('*degree*C*')'),yaxs='i',xaxs='i')
> axis(1,at=c(colow.F,cohigh.F),labels=as.character(c(colow.F,cohigh.F)))
> legend('bottomright',as.character(sort(unique(Flength))),
+       pch=19, col=colL[sort(unique(Flength))-11],bg='white')

```

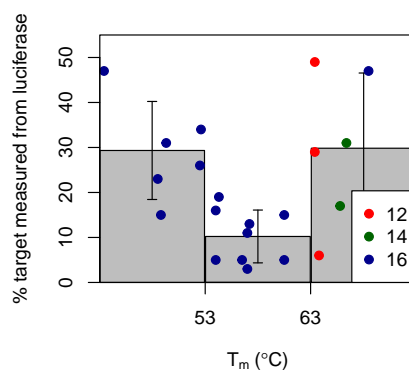


Figure 2d: 21 oligonucleotides targeted against the luciferase firefly gene.

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

```

> ##### We plot the data from Stanton et al 2012
> dat.S <- dat[dat[,1]=="Stanton 2012",]
> cohigh.S <- 61; colow.S <- 46
> tmp <- abs(cohigh.S-colow.S)
> cut.S <- cut(dat.S$Predicted.Tm,c(0,colow.S,cohigh.S,100),labels=F)
> Sx <- lapply(1:3,function(i)dat.S$Predicted.Tm[cut.S==i])
> Sy <- lapply(1:3,function(i) dat.S$Dose.3nm[cut.S==i])
> Slength <- dat.S$Oligo.length
> bp <- barplot(sapply(Sy,mean),ylim=c(0,105), las=1,axes=F,

```

```

+           yaxs='i', xaxs='i',space=0.01)
> plotCI(bp[,1],sapply(Sy,mean),sapply(Sy,sd),add=T,pch=NA, gap=0,yaxs='i')
> par(new=T)
> plot(unlist(Sx),unlist(Sy),xlim=c(colow.S-tmp,cohigh.S+tmp),
+       ylim=c(0,105), pch=19,col=colL[Slength-11],xaxt='n',
+       ylab='% target measured from PCR',
+       xlab=expression(T[m]~'(*degree*C*')'),yaxs='i',xaxs='i')
> axis(1,at=c(colow.S,cohigh.S),labels=as.character(c(colow.S,cohigh.S)))
> legend('bottomright',as.character(sort(unique(Slength))),
+       pch=19,col=colL[sort(unique(Slength))-11],bg='white')

```

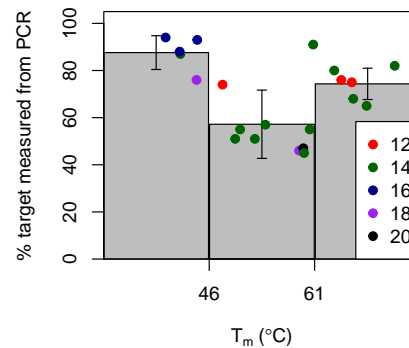


Figure 2e: 21 oligonucleotides targeted against the glucocorticoid receptor.

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

```

> ### We plot the data from Pedersen et al, 2013
> dat.P <- dat[dat$Study=="Pedersen 2013",]
> cohigh.P <- 56; colow.P <- 47
> tmp <- abs(cohigh.P-colow.P)
> cut.P <- cut(dat.P$Predicted.Tm,c(0,colow.P,cohigh.P,100),labels=F)
> Px <- lapply(1:3,function(i)dat.P$Predicted.Tm[cut.P==i])
> Py <- lapply(1:3,function(i) dat.P$IC50[cut.P==i])
> Plength <- dat.P$Oligo.length
> bp <- barplot(sapply(Py,mean),ylim=c(0,0.016), las=1,
+               axes=F,yaxs='i', xaxs='i',space=0.01)
> plotCI(bp[,1],sapply(Py,mean),sapply(Py,sd),add=T,
+         pch=NA, gap=0,yaxs='i')
> par(new=T)
> plot(unlist(Px),unlist(Py),xlim=c(colow.P-tmp,cohigh.P+tmp),ylim=c(0,0.016),

```

```

+     pch=19,col=colL[Plength-11],xaxt='n',ylab=expression(IC[50]~'('*nM*')'),
+     xlab=expression(T[m]~'('*degree*C*')'),yaxs='i',xaxs='i')
> axis(1,at=c(colow.P,cohigh.P),labels=as.character(c(colow.P,cohigh.P)))
> legend('bottomright',as.character(sort(unique(Plength))),
+       pch=19,col=colL[sort(unique(Plength))-11],bg='white')

```

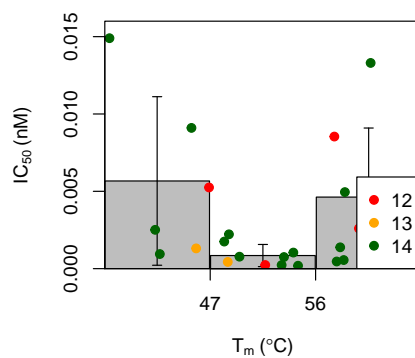


Figure 2f: 23 oligonucleotides targeted against ApoB.