

Figure code for 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 functions from the ASOmodels 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])))

```

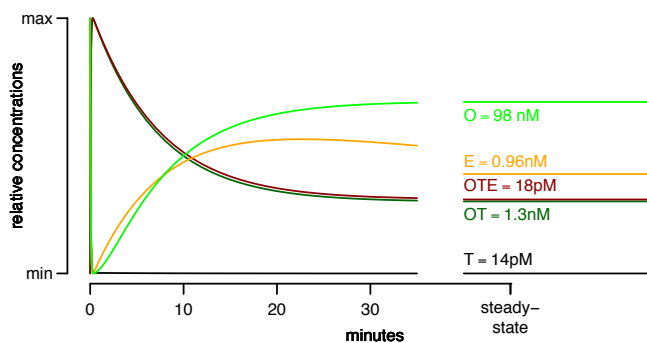


Figure 2a: Time resolved simulation of the relative concentrations of key species

Figure 2b: Simulated dose-response curve

Given a set of parameters the 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)
> abline(v=IC50(parms['KdOT']),lty=2)
> 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)
```

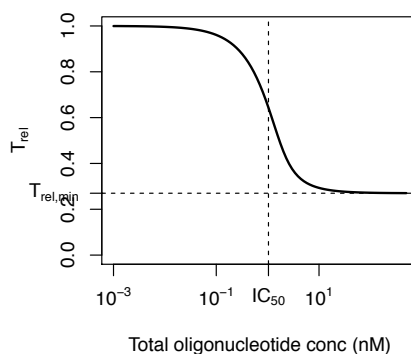


Figure 2b: The relative total target concentration (T_{rel}) is defined as the steady state level of total target in the presence of oligonucleotide divided by the target concentration in the absence of oligonucleotide. Dashed lines indicate efficacy (horizontal) and IC_{50} (vertical)

Figure 2c: An optimum affinity

For a sequence of affinities `D1_seq` the IC_{50} values are calculated by use of the 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)

```

For the sequence of affinities the two different 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')

```

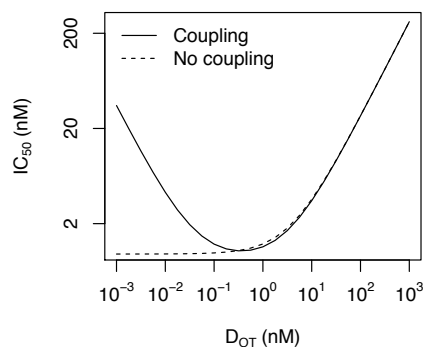


Figure 2c: The IC_{50} as a function of the dissociation constant for the OT complex. A low K_{dOT} corresponds to a high affinity binding. Dashed line: no coupling of off-rates. Solid line: coupling of off-rates.

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$Predicted.Tm,c(0,colow.F,cohigh.F,100),labels=F)
> Fx <- lapply(1:3,function(i)dat.F$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')

```

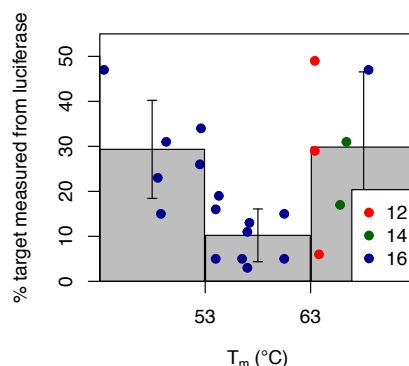


Figure 2d: 21 oligonucleotides targeted against the luciferase firefly gene.

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')

```

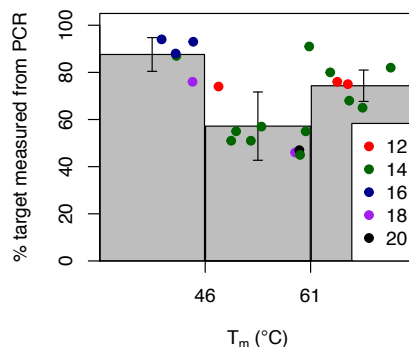


Figure 2e: 21 oligonucleotides targeted against the glucocorticoid receptor.

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)

```

Figure 2 is a scatter plot showing the relationship between the melting temperature (T_m) and the inhibitory concentration (IC_{50}) for three different conditions (12, 13, and 14). The x-axis represents T_m in degrees Celsius ($^{\circ}C$), with major ticks at 47 and 56. The y-axis represents IC_{50} in nanomolar (nM), with major ticks at 0.000, 0.005, 0.010, and 0.015. The data points are color-coded: red for 12, yellow for 13, and green for 14. Shaded gray regions indicate the distribution of T_m values for each series. Error bars are shown for each data point.

Series	T_m ($^{\circ}C$)	IC_{50} (nM)
12	~48	~0.0045
12	~56	~0.0085
13	~47	~0.0005
13	~48	~0.0005
14	~45	~0.0145
14	~46	~0.0025
14	~47	~0.0005
14	~48	~0.0015
14	~49	~0.0005
14	~50	~0.0005
14	~51	~0.0005
14	~52	~0.0005
14	~53	~0.0005
14	~54	~0.0005
14	~55	~0.0005
14	~56	~0.0005
14	~57	~0.0005
14	~58	~0.0005
14	~59	~0.0005
14	~60	~0.0005
14	~61	~0.0005
14	~62	~0.0005
14	~63	~0.0005
14	~64	~0.0005
14	~65	~0.0005
14	~66	~0.0005
14	~67	~0.0005
14	~68	~0.0005
14	~69	~0.0005
14	~70	~0.0005
14	~71	~0.0005
14	~72	~0.0005
14	~73	~0.0005
14	~74	~0.0005
14	~75	~0.0005
14	~76	~0.0005
14	~77	~0.0005
14	~78	~0.0005
14	~79	~0.0005
14	~80	~0.0005
14	~81	~0.0005
14	~82	~0.0005
14	~83	~0.0005
14	~84	~0.0005
14	~85	~0.0005
14	~86	~0.0005
14	~87	~0.0005
14	~88	~0.0005
14	~89	~0.0005
14	~90	~0.0005
14	~91	~0.0005
14	~92	~0.0005
14	~93	~0.0005
14	~94	~0.0005
14	~95	~0.0005
14	~96	~0.0005
14	~97	~0.0005
14	~98	~0.0005
14	~99	~0.0005
14	~100	~0.0005

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