

# TAC linearity summary

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11/12/22

#clear R env

```
rm(list = ls())
```

Load libraries

```
library(ggplot2)
library(ggpmisc)
library(plyr)
library(dplyr)
library(tidyr)
```

Import data file

```
TAC<-read.csv("sample_data.csv", na.strings=c("", "NA"))
```

Plot a standard curve and summarize efficiency for each target

```
#Format data and add column with logs of quantities
TAC$CT<-as.numeric(gsub(",", "", TAC$CT))

TAC$Quantity<-as.numeric(gsub(",", "", TAC$Quantity))
TAC$Quantity_log<-log10(TAC$Quantity)

#summarize efficiencies
TAC<-TAC[!is.na(TAC$CT),]
TAC<-TAC[!is.na(TAC$Quantity_log),]
TAC<-TAC[rowSums(is.na(TAC)) != ncol(TAC), ]
efficiency_summ<-ddply(TAC, "Target.Name", function(x) {
  model <- lm(CT ~ Quantity_log, data = x)
  coef(model)})

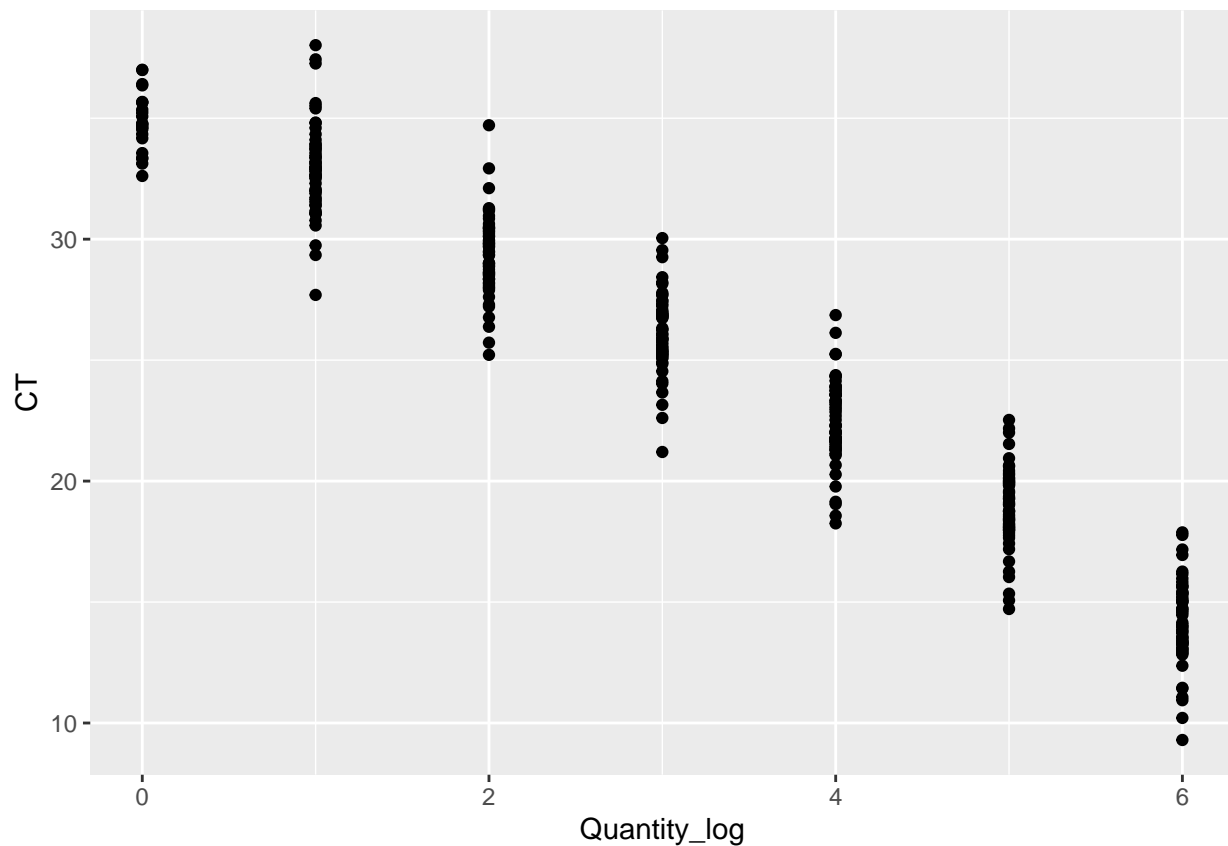
colnames(efficiency_summ) <- c("Target.Name", "Y.intercept", "Slope")

efficiency_summ$Efficiency<-10^(-1/efficiency_summ$Slope)-1

#create a list of dataframes by target
targets=by(TAC, TAC[, "Target.Name"], function(x) x)

#plot of standard curve points across all targets
```

```
p<-ggplot(TAC, aes(x = Quantity_log, y = CT)) +
  geom_point()
p
```



```
#plot standard curves for all targets
target_plots <- function(targets){
  ggplot(targets, aes(x = Quantity_log, y = CT)) +
    geom_point()+
    stat_poly_line(formula = y~x, se=FALSE) +
    stat_poly_eq(aes(label = paste(after_stat(eq.label), after_stat(rr.label), sep = "*\\", \\"*\")), formula = y~x)
}

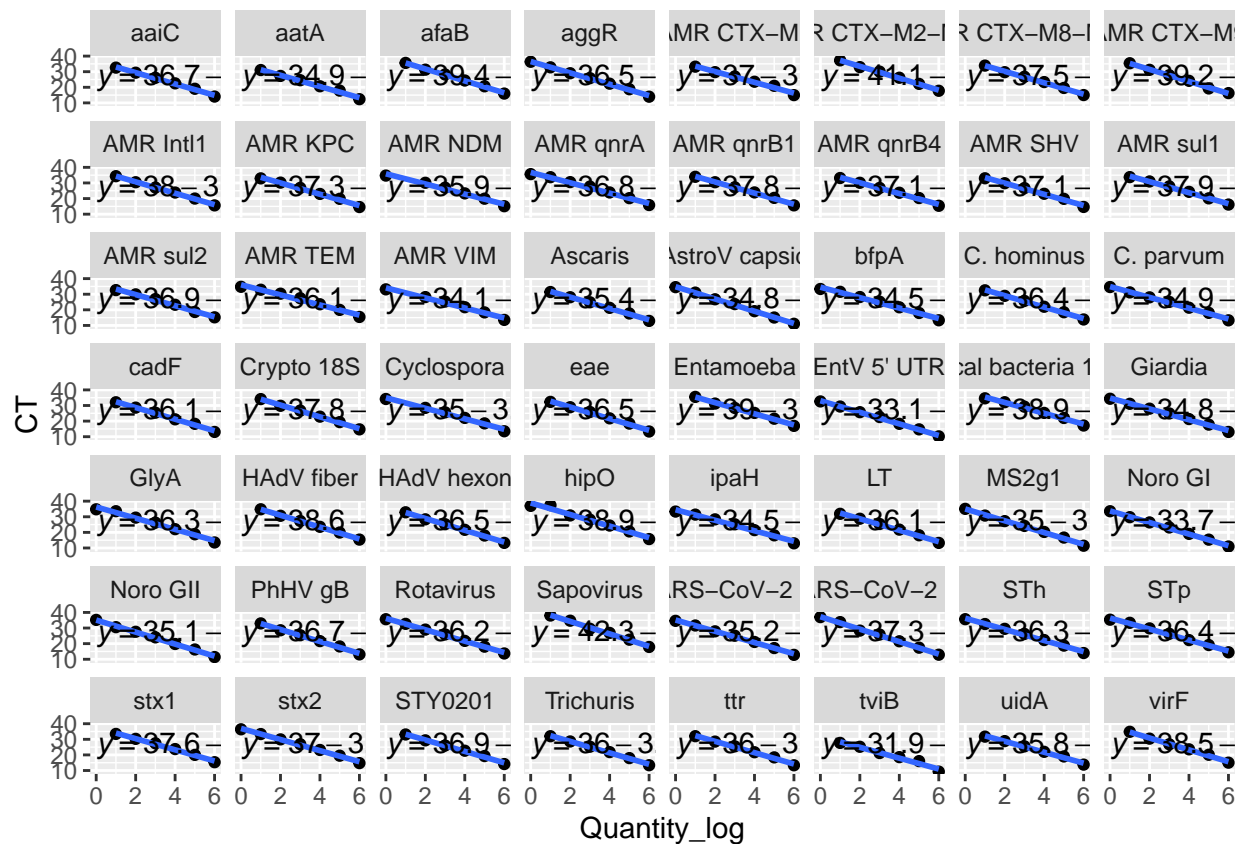
q1 <- lapply(targets, target_plots)

q2 <- lapply(seq_along(q1), function(i) {
  q1[[i]] + ggtitle(names(targets)[i])
})

#q2 #careful--this command makes a lot of plots!

#faceted plot with all adjusted standard curves for all targets
r<-ggplot(TAC, aes(x = Quantity_log, y = CT)) +
  geom_point() +
  facet_wrap(~ Target.Name)+
  stat_poly_line(formula = y~x, se=FALSE) +
```

```
stat_poly_eq(aes(label = paste(after_stat(eq.label), after_stat(rr.label), sep = "*\\", \\"*")), formula = y ~ x)
```



Plot standard curves and summarize efficiencies without last point in the curve (where it flattens out as it approaches the LOD)

```
#remove last point on standard curve
TAC_adj <- TAC[TAC$Quantity_log != 0, ]
#TAC_adj <- TAC[TAC$Quantity_log != 0, ] #run this to look at standard curves without last two dilution

#create a list of dataframes by target
targets_adj=by(TAC_adj, TAC_adj[, "Target.Name"], function(x) x)

#plot adjusted standard curves for all targets
target_plots_adj <- function(targets_adj){
  ggplot(targets_adj, aes(x = Quantity_log, y = CT)) +
    geom_point()+
    stat_poly_line(formula = y~x, se=FALSE) +
    stat_poly_eq(aes(label = paste(after_stat(eq.label), after_stat(rr.label), sep = "*\\", \\"*")), formula = y ~ x)
}

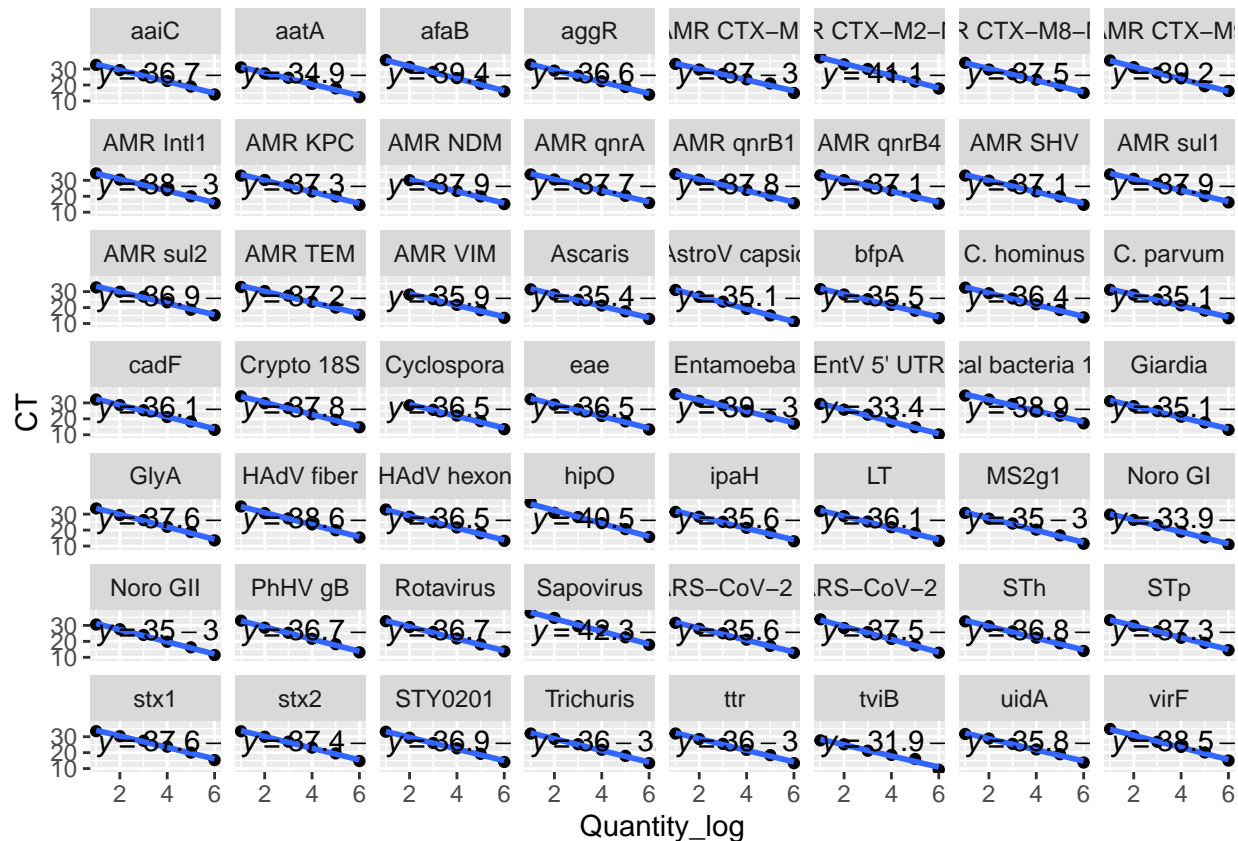
s1 <- lapply(targets_adj, target_plots_adj)

s2 <- lapply(seq_along(s1), function(i) {
  s1[[i]] + ggtitle(names(targets_adj)[i])
})
```

*#s2 #careful-this command makes a lot of plots!*

*#faceted plot with all adjusted standard curves for all targets*

```
t<-ggplot(TAC_adj, aes(x = Quantity_log, y = CT)) +
  geom_point() +
  facet_wrap(~ Target.Name)+
  stat_poly_line(formula = y~x, se=FALSE) +
  stat_poly_eq(aes(label = paste(after_stat(eq.label), after_stat(rr.label), sep = "*\\", \\"*\")), formula = y~x, se=FALSE)
```



Summarize dilutions with detection and high and low CT values for each target FIX THIS

```
TAC_CT<-TAC %>% select(Target.Name, Sample.Name, CT, R.superscript.2.)
```

```
high_low<-TAC_CT %>%
  group_by(Target.Name) %>%
  mutate(
    CT_min=min(CT, na.rm=T),
    CT_max=max(CT, na.rm=T)) %>%
  mutate(
    Dilution_min=min(Sample.Name, na.rm=T),
    Dilution_max=max(Sample.Name, na.rm=T)) %>%
  select(-Sample.Name) %>%
  select(-CT)
```

Summarize adjusted (without lowest dilution) slope, efficiency, and R2 values for each target

```

R2<- TAC_adj %>%
  group_by(Target.Name) %>%
  dplyr::summarise(model = list(lm(CT ~ Quantity_log, data = cur_data()))),
  coef = list(coef(model[[1]])),
  R2_adj = summary(model[[1]])$r.sq)%>%
  unnest_wider(coef, names_repair = 'unique')

R2$Efficiency_adj<- 10^(-1/R2$Quantity_log)-1

```

Summarize NTC results

```

TAC_NTC<-read.csv("sample_data.csv", na.strings=c("", "NA"))

TAC_NTC <- TAC_NTC %>%
  subset(Sample.Name=="NTC")

TAC_NTC$CT<-as.numeric(gsub(",", "", TAC_NTC$CT))
TAC_NTC<- TAC_NTC%>%
  select(Target.Name, CT)%>%
  dplyr::rename(NTC_CT=CT)

```

Summary table

```

summary<-merge(efficiency_summ, R2)
summary<-merge(high_low, summary)
summary<-merge(TAC_NTC, summary)
summary <- summary %>%
  distinct(.keep_all = TRUE) %>%
  select(-model) %>%
  relocate(R.superscript.2., .before=Efficiency) %>%
  relocate(NTC_CT, .before=Y.intercept)%>%
  dplyr::rename(R2=R.superscript.2.) %>%
  dplyr::rename(Y.intercept_adj='(Intercept)') %>%
  dplyr::rename(Slope_adj=Quantity_log)%>%
  mutate_if(is.numeric, round, digits=2)

summary

```

##	Target.Name	CT_min	CT_max	Dilution_min	Dilution_max	NTC_CT
## 1	aaiC	14.07	32.61	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 2	aatA	12.37	31.06	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 3	afaB	15.97	35.63	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 4	aggR	13.97	36.36	PCP 10 <sup>0</sup>	PCP 10 <sup>6</sup>	NA
## 5	AMR CTX-M1	14.99	33.21	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 6	AMR CTX-M2-M74	17.77	37.26	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 7	AMR CTX-M8-M25	15.03	33.90	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 8	AMR CTX-M9	16.26	35.41	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 9	AMR Int11	15.62	34.34	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 10	AMR KPC	14.57	32.94	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA
## 11	AMR NDM	15.10	34.74	PCP 10 <sup>0</sup>	PCP 10 <sup>6</sup>	NA
## 12	AMR qnrA	15.82	35.65	PCP 10 <sup>0</sup>	PCP 10 <sup>6</sup>	NA
## 13	AMR qnrB1	15.64	33.94	PCP 10 <sup>1</sup>	PCP 10 <sup>6</sup>	NA

## 14	AMR qnrB4	15.37	33.16	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 15	AMR SHV	14.64	33.00	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 16	AMR sul1	16.16	33.74	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 17	AMR sul2	15.20	32.64	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 18	AMR TEM	15.43	34.76	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 19	AMR VIM	13.55	33.12	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 20	Ascaris	12.90	31.42	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 21	AstroV capsid	11.07	34.54	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 22	bfpA	13.32	33.36	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 23	C. hominus	13.84	32.51	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 24	C. parvum	13.27	34.62	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 25	cadF	13.05	32.07	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 26	Crypto 18S	14.72	34.11	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 27	Cyclospora	13.50	34.17	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 28	eae	13.35	32.31	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 29	Entamoeba	16.95	35.51	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 30	EntV 5' UTR	10.21	32.61	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 31	Fecal bacteria 16S	17.17	34.60	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 32	Giardia	13.05	34.34	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 33	GlyA	13.55	34.79	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 34	HAdV fiber	15.37	34.80	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 35	HAdV hexon	13.26	32.87	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 36	hipO	15.69	37.43	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 37	ipaH	12.98	33.33	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 38	LT	13.26	31.89	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 39	MS2g1	11.46	35.07	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 40	Noro GII	11.43	35.23	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 41	PhHV gB	13.12	32.97	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 42	Rotavirus	13.71	35.65	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 43	Sapovirus	17.88	38.02	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 44	SARS-CoV-2 N1	12.83	34.63	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	10.66
## 45	SARS-CoV-2 N2	12.91	37.00	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	15.00
## 46	STh	13.98	35.67	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 47	STp	14.46	35.35	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 48	stx1	15.32	33.46	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 49	stx2	14.74	36.41	PCP 10 <sup>-0</sup>	PCP 10 <sup>-6</sup>	NA
## 50	STY0201	14.15	33.11	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 51	Trichuris	13.40	31.99	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 52	ttr	13.37	32.00	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 53	tviB	9.30	27.70	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 54	uidA	13.79	31.71	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
## 55	virF	15.04	34.82	PCP 10 <sup>-1</sup>	PCP 10 <sup>-6</sup>	NA
##	Y.intercept Slope	R2	Efficiency	Y.intercept_adj	Slope_adj	R2_adj
## 1	36.67 -3.63	0.99	0.88	36.67	-3.63	0.99
## 2	34.86 -3.59	0.99	0.90	34.86	-3.59	0.99
## 3	39.42 -3.83	1.00	0.82	39.42	-3.83	1.00
## 4	36.50 -3.64	1.00	0.88	36.63	-3.67	1.00
## 5	36.98 -3.45	0.98	0.95	36.98	-3.45	0.98
## 6	41.08 -3.82	1.00	0.83	41.08	-3.82	1.00
## 7	37.53 -3.66	1.00	0.88	37.53	-3.66	1.00
## 8	39.17 -3.86	1.00	0.82	39.17	-3.86	1.00
## 9	38.04 -3.66	1.00	0.88	38.04	-3.66	1.00
## 10	37.26 -3.63	0.99	0.89	37.26	-3.63	0.99
## 11	35.88 -3.26	0.98	1.03	37.94	-3.71	0.99

## 12	36.75	-3.32	0.99	1.00	37.71	-3.54	0.99
## 13	37.77	-3.57	1.00	0.91	37.77	-3.57	1.00
## 14	37.06	-3.47	0.99	0.94	37.06	-3.47	0.99
## 15	37.06	-3.59	0.99	0.90	37.06	-3.59	0.99
## 16	37.91	-3.54	0.99	0.92	37.91	-3.54	0.99
## 17	36.90	-3.57	0.99	0.91	36.90	-3.57	0.99
## 18	36.05	-3.24	0.98	1.04	37.18	-3.50	0.99
## 19	34.11	-3.22	0.98	1.04	35.89	-3.62	0.99
## 20	35.44	-3.64	1.00	0.88	35.44	-3.64	1.00
## 21	34.83	-3.93	1.00	0.80	35.08	-3.99	1.00
## 22	34.52	-3.35	0.99	0.99	35.52	-3.58	0.99
## 23	36.45	-3.68	1.00	0.87	36.45	-3.68	1.00
## 24	34.89	-3.46	1.00	0.95	35.13	-3.52	0.99
## 25	36.14	-3.73	0.99	0.85	36.14	-3.73	0.99
## 26	37.83	-3.78	1.00	0.84	37.83	-3.78	1.00
## 27	34.99	-3.38	0.99	0.98	36.47	-3.71	0.99
## 28	36.47	-3.73	0.99	0.86	36.47	-3.73	0.99
## 29	38.99	-3.57	0.99	0.90	38.99	-3.57	0.99
## 30	33.06	-3.71	1.00	0.86	33.45	-3.80	1.00
## 31	38.88	-3.47	0.99	0.94	38.88	-3.47	0.99
## 32	34.76	-3.47	1.00	0.94	35.13	-3.56	0.99
## 33	36.28	-3.60	0.98	0.90	37.58	-3.90	1.00
## 34	38.58	-3.80	1.00	0.83	38.58	-3.80	1.00
## 35	36.50	-3.79	1.00	0.84	36.50	-3.79	1.00
## 36	38.88	-3.72	0.97	0.86	40.51	-4.09	0.99
## 37	34.52	-3.38	0.99	0.98	35.56	-3.61	0.99
## 38	36.05	-3.67	0.99	0.87	36.05	-3.67	0.99
## 39	35.01	-3.79	1.00	0.84	34.96	-3.77	0.99
## 40	35.11	-3.85	1.00	0.82	35.01	-3.83	1.00
## 41	36.73	-3.83	1.00	0.82	36.73	-3.83	1.00
## 42	36.22	-3.65	1.00	0.88	36.70	-3.76	1.00
## 43	42.25	-4.00	1.00	0.78	42.25	-4.00	1.00
## 44	35.17	-3.61	1.00	0.89	35.64	-3.72	1.00
## 45	37.26	-4.00	1.00	0.78	37.48	-4.05	0.99
## 46	36.28	-3.57	0.99	0.91	36.81	-3.69	0.99
## 47	36.38	-3.52	0.99	0.92	37.27	-3.73	1.00
## 48	37.57	-3.58	0.99	0.90	37.57	-3.58	0.99
## 49	36.96	-3.56	1.00	0.91	37.44	-3.66	1.00
## 50	36.94	-3.67	0.99	0.87	36.94	-3.67	0.99
## 51	36.02	-3.67	1.00	0.87	36.02	-3.67	1.00
## 52	36.00	-3.64	0.99	0.88	36.00	-3.64	0.99
## 53	31.89	-3.49	0.97	0.93	31.89	-3.49	0.97
## 54	35.75	-3.51	0.99	0.93	35.75	-3.51	0.99
## 55	38.45	-3.81	1.00	0.83	38.45	-3.81	1.00
##	Efficiency_adj						
## 1	0.88						
## 2	0.90						
## 3	0.82						
## 4	0.87						
## 5	0.95						
## 6	0.83						
## 7	0.88						
## 8	0.82						
## 9	0.88						

## 10	0.89
## 11	0.86
## 12	0.92
## 13	0.91
## 14	0.94
## 15	0.90
## 16	0.92
## 17	0.91
## 18	0.93
## 19	0.89
## 20	0.88
## 21	0.78
## 22	0.90
## 23	0.87
## 24	0.93
## 25	0.85
## 26	0.84
## 27	0.86
## 28	0.86
## 29	0.90
## 30	0.83
## 31	0.94
## 32	0.91
## 33	0.81
## 34	0.83
## 35	0.84
## 36	0.76
## 37	0.89
## 38	0.87
## 39	0.84
## 40	0.82
## 41	0.82
## 42	0.84
## 43	0.78
## 44	0.86
## 45	0.77
## 46	0.87
## 47	0.85
## 48	0.90
## 49	0.87
## 50	0.87
## 51	0.87
## 52	0.88
## 53	0.93
## 54	0.93
## 55	0.83

```
write.csv(summary, file="TAC_sample_linearity_summary.csv")
```