```
library(knitr)
```

```
data <- read.csv("uval.csv")</pre>
```

### Question 1

Table 1: Coefficients & Std. Error of linear model

	Estimate	Std. Error	t value	$\Pr(> t )$
(Intercept)	-0.03525	0.00665	-5.30037	0.00000
underval	0.00476	0.00218	2.18614	0.02898
log(gdp)	0.00630	0.00079	7.96591	0.00000

Since the coefficient of log(gdp) is positive, this model does not seem to support the idea of "catching-up" as countries with higher GDP have a higher economic growth rate. However, it does support the idea that under-valuing a currency boosts economic growth as the coefficient of underval is positive, indicating a positive underval index, which represents undervaluing, leads to higher economic growth.

# Question 2

Q2 a)

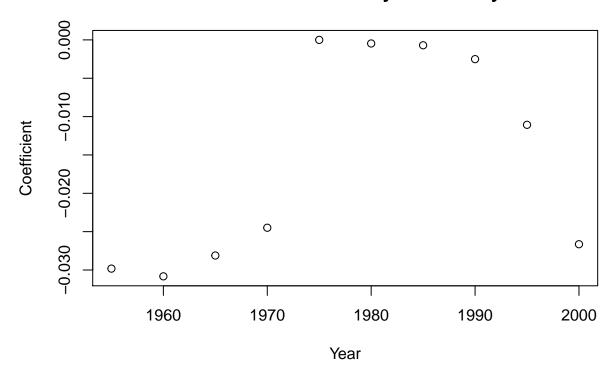
Table 2: Coefficients & Std. Error of linear model

	Estimate	Std. Error	t value	Pr(> t )
underval	0.01361	0.00290	4.69667	0
$\log(\mathrm{gdp})$	0.02892	0.00317	9.13254	0

**Q2 b)** It is more appropriate to use factor(year) as there are only unique years that are 5 years apart. As such, modelling this way we will have a slope for each 5-year interval rather than a single slope for each increment of year.

Q2 c)

## Plot of coefficient of each year across years



### Q2 d)

Since the coefficient of log(gdp) is positive, this model does not seem to support the idea of "catching-up" as countries with higher GDP have a higher economic growth rate. However, it does support the idea that under-valuing a currency boosts economic growth as the coefficient of underval is positive, indicating a positive underval index, which represents undervaluing, leads to higher economic growth.

## Question 3

#### Q3 a)

Table 3:  $\mathbb{R}^2$  values for each linear model

	Model 1	Model 2
$R^2$	0.04855	0.42924
Adj. $R^2$	0.04709	0.33214

#### Q3 b)

```
# Taken from textbook chapter 3 page 77
cv.lm <- function(data, formulae, nfolds = 5) {</pre>
    data <- na.omit(data)</pre>
    formulae <- sapply(formulae, as.formula)</pre>
    n <- nrow(data)</pre>
    fold.labels <- sample(rep(1:nfolds, length.out = n))</pre>
    mses <- matrix(NA, nrow = nfolds, ncol = length(formulae))</pre>
    colnames <- as.character(formulae)</pre>
    for (fold in 1:nfolds) {
        test.rows <- which(fold.labels == fold)</pre>
        train <- data[-test.rows, ]</pre>
        test <- data[test.rows, ]</pre>
        for (form in 1:length(formulae)) {
             current.model <- lm(formula = formulae[[form]], data = train)</pre>
             predictions <- predict(current.model, newdata = test)</pre>
             test.responses <- eval(formulae[[form]][[2]], envir = test)</pre>
             test.errors <- test.responses - predictions</pre>
             mses[fold, form] <- mean(test.errors^2)</pre>
        }
    }
    return(colMeans(mses))
loocv.mse <- cv.lm(data, c("growth ~ underval + log(gdp)", "growth ~ underval + log(gdp) + factor(count
```

```
loocv.mse <- cv.lm(data, c("growth ~ underval + log(gdp)", "growth ~ underval + log(gdp) + r
names(loocv.mse) <- c("Model 1", "Model 2")
kable(loocv.mse, caption = "$\\hat{MSE}$ of linear models by LOOCV")</pre>
```

Table 4:  $\hat{MSE}$  of linear models by LOOCV

2
0.001030348892
0.000952766927