1. **Problem**
   Theory: Consider a linear regression of `y` on `x`. It is usually estimated with which estimation technique (three-letter abbreviation)?

   This estimator yields the best linear unbiased estimator (BLUE) under the assumptions of the Gauss-Markov theorem. Which of the following properties are required for the errors of the linear regression model under these assumptions?

   independent / zero expectation / normally distributed / identically distributed / homoscedastic

   Application: Using the data provided in `linreg.csv` estimate a linear regression of `y` on `x`. What are the estimated parameters?

   Intercept:

   Slope:

   In terms of significance at 5% level:

   `x` and `y` are not significantly correlated / `y` increases significantly with `x` / `y` decreases significantly with `x`

   Solution
   Theory: Linear regression models are typically estimated by ordinary least squares (OLS). The Gauss-Markov theorem establishes certain optimality properties: Namely, if the errors have expectation zero, constant variance (homoscedastic), no autocorrelation and the regressors are exogenous and not linearly dependent, the OLS estimator is the best linear unbiased estimator (BLUE).

   Application: The estimated coefficients along with their significances are reported in the summary of the fitted regression model, showing that `y` increases significantly with `x` (at 5% level).

```
Call:
lm(formula = y ~ x, data = d)

Residuals:
     Min       1Q   Median       3Q      Max 
-0.50503 -0.17149 -0.01047  0.13726  0.69840 

Coefficients:
             Estimate Std. Error t value Pr(>|t|)    
(Intercept) -0.005094   0.023993  -0.212    0.832    
x            0.558063   0.044927  12.421   <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2399 on 98 degrees of freedom
Multiple R-squared:  0.6116,	Adjusted R-squared:  0.6076 
F-statistic: 154.3 on 1 and 98 DF,  p-value: < 2.2e-16
```

   Code: The analysis can be replicated in R using the following code.

```
## data
d <- read.csv("linreg.csv")
## regression
m <- lm(y ~ x, data = d)
summary(m)
## visualization
```

```
plot(y ~ x, data = d)
abline(m)
```