

# Introduction to Statistic for Data Science

## Group Mini-Project Presentation: Happiness Ladder

### ISDS Group 10

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- 1 Introduction
  - Intro to Data Set and its Context

- 2 Statistical Modeling
  - Model Explanation
  - Model Construction

- 3 Conclusions and Analysis
  - Results and Analysis
  - Conclusions

- 4 Future Work

## 1 Introduction

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- Model Construction

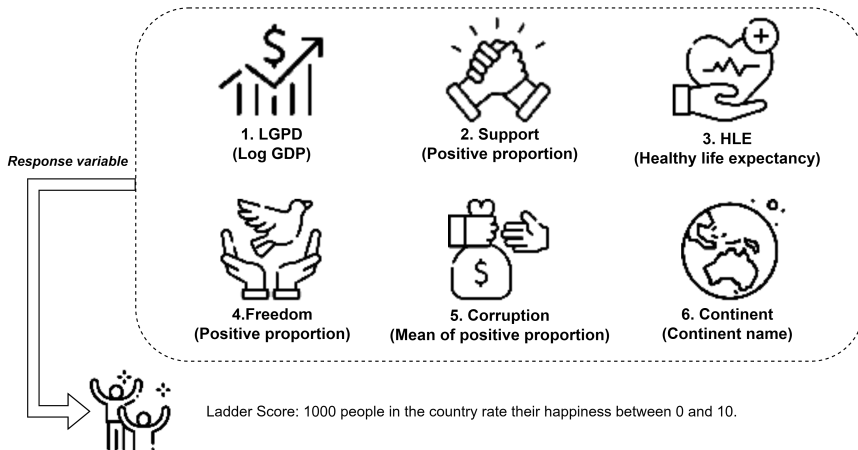
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# Data Set and its Context

## “Happiness Ladder” — Data collected from 137 countries



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# Multiple Linear Regression Model

- **Model Target:** Some socio-economic indexes are used to predict Ladder Score to assist government decision-making.
- **Independent Variables:** LGDP, Support, HLE, Freedom, Corruption, Continent
- **Dependent Variables:** Ladder Score
- **Model Function:**  
$$\text{input}(\text{LGDP}, \text{Support}, \dots) \Rightarrow \text{output}(\text{Ladder Score})$$
$$\text{LadderScore} = \beta_0 + \beta_1 * \text{LGDP} + \beta_2 * \text{Support} + \dots + \epsilon$$
- **Optimization Target:** Making the model with the **best subset** and **higher Adjusted R-squared**.  
$$\begin{cases} \text{Select the independent variables} \\ \text{Update } \beta \text{ and } \epsilon \text{ to minimize the residual sum of squared} \end{cases}$$

# Why we choose Multiple Linear Regression Model

Aspect	Simple Linear Regression	Multiple Linear Regression
Model	$Y = \beta_0 + \beta_1 X + \varepsilon$	$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k + \varepsilon$
Advantages	Simple and easy interpretation	Captures complex relationships with multiple predictors
	Suitable for examining two-variable relationships	Considers multiple factors, offering a comprehensive view
	Less prone to overfitting with fewer predictors	Analyzes independent effects of each predictor
Disadvantages	Limited to two-variable relationships	More complex, challenging interpretation
	Assumes a linear relationship	Susceptible to multicollinearity with correlated predictors
	May not capture real-world complexity	More assumptions (linearity, independence, normality)
		Risk of overfitting, especially with many predictors

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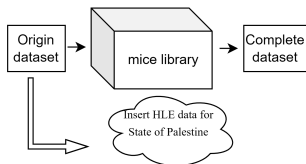
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# Constructing the Model – Data Processing

## Data Pre-process

### Missing Value Handling



### Country area error

Fix it with our common sense knowledge

Ecudor → South America

Uganda → Africa

## Dataset Segmentation

Continent Mapping

"North America" = 1  
"South America" = 2  
"Europe" = 3  
"Asia" = 4  
"Africa" = 5  
"Oceania" = 6

Serperate Dataset according to Continent

### Our Dataset

Happy\_general

Happy\_general\_continent

Africa

Asia

Europe

...

...

# Constructing the Model – Dataset Inspection

Dataset Name	Country_name	LGDP	Support	HLE	Freedom	Corruption	Continent	Numeric Continent	Ladder_score
Happy_origin	√	√	√	with NA	√	√	√		√
Happy_complete	√	√	√	√	√	√	√		√
Happy_general_continent		√	√	√	√	√	√	√	√
Happy_general		√	√	√	√	√	√		√
Africa		√	√	√	√	√	Africa		√
Asia		√	√	√	√	√	Asia		√
Europe		√	√	√	√	√	Europe		√
North_America		√	√	√	√	√	North_America		√
Oceania		√	√	√	√	√	Oceania		√
South_America		√	√	√	√	√	South_America		√

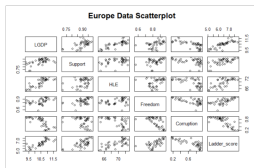
# Constructing the Model – Bring dataset to MLR model

## Step 1: Correlationship

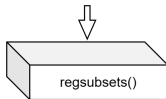
	LGDP	Support	HLE	Freedom	Corruption	Ladder_score
LGDP	1.00000000	0.52707883	0.54785932	0.009994757	0.07777944	0.4917032
Support	0.52707883	1.00000000	0.23504723	0.074828973	0.26647977	0.4007134
HLE	0.54785932	0.23504723	1.00000000	-0.191237780	-0.09175306	0.2603294
Freedom	0.009994757	0.07482897	-0.19123778	1.000000000	-0.14779236	0.1954914
Corruption	0.07777944	0.26647977	-0.09175306	-0.14779236	1.000000000	0.0661934
Ladder_score	0.4917032	0.4007134	0.2603294	0.1954914	0.0661934	1.0000000

	LGDP	Support	HLE	Freedom	Corruption	Ladder_score
LGDP	1.00000000	0.5278129	0.77461875	0.34362034	-0.4851374	0.6519759
Support	0.5278129	1.00000000	0.52751967	0.446975919	-0.2252249	0.8556972
HLE	0.7746188	0.5275197	1.00000000	0.96947368	-0.4939959	0.5087794
Freedom	0.3436201	0.44697592	0.96947368	1.000000000	-0.2056149	0.6120383
Corruption	-0.4851974	-0.2252249	-0.4939959	-0.20561492	1.00000000	-0.3589953
Ladder_score	0.6519759	0.8556972	0.50877939	0.61203835	-0.3589953	1.0000000

## Step 2: Scatterplot

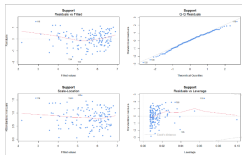


## Step 3: Finding the best subset



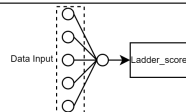
	LGDP	Support	HLE	Freedom	Corruption
1	( 1 )	''	''	''	''
2	( 1 )	''	''	''	''
3	( 1 )	''	''	''	''
4	( 1 )	''	''	''	''
5	( 1 )	''	''	''	''

## Step 4: Compared with Single Linear Regression



## Step 5: Regression Analytics

### Calculate the model



### Model Evaluation

Variance Inflation Factor Check

Autocorrelation -- Residuals

Adjust R squared

### Future Analytics

Interaction between variables

non-linear transformation of predictors

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# Model Results and Analysis

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# Conclusions and Insights

# Next Steps