

Predicting Age with Multiple Linear Regression

Agenda

Introduction

Hypothesis

Data-Analysis Process

Data Visualizations

Findings

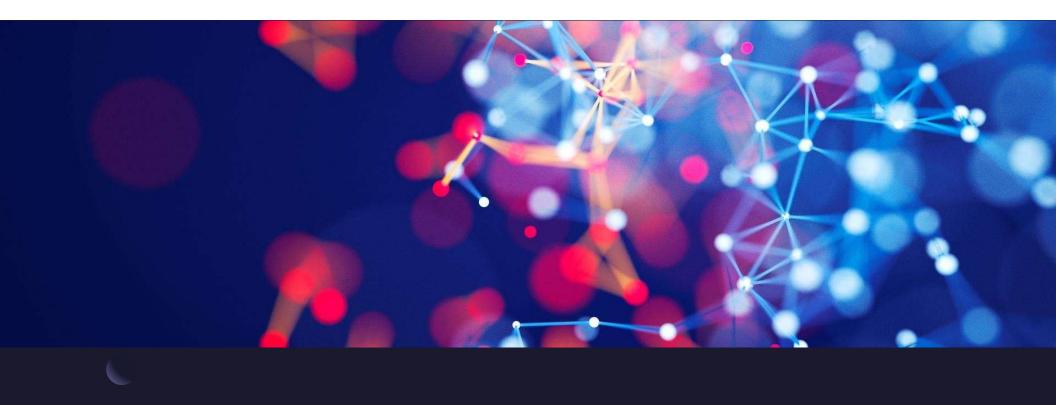
Limitations

Summary

Introduction

- Kim Fowler
- Completing a Master's degree in Data Analytics from WGU
- Database Developer with 10 years of experience
- Predicting a person's age (years) from a set of health and lifestyle variables





Hypothesis

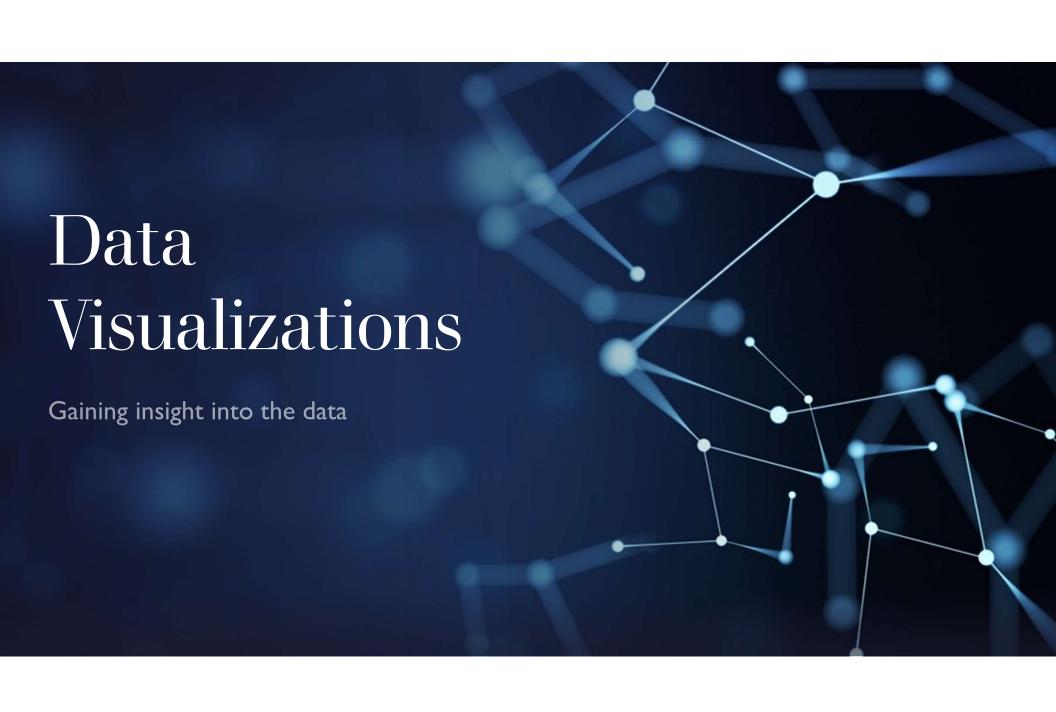


- Hypothesis The variables Cholesterol Level, Bone Density, Blood Glucose Level, Vision Sharpness, Hearing Ability, and Smoking Status do impact Age (years).
- Alterative Hypothesis The variables Cholesterol Level, Bone Density, Blood Glucose Level, Vision Sharpness, Hearing Ability, and Smoking Status do not impact Age (years).

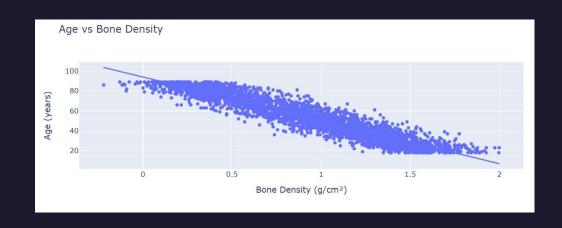
Data-Analysis Process

- Data Collection
- Exploratory Data Analysis
- Prepare the Data
- Data Wrangling
- Predictive Modeling
- Feature Selection
- Model Evaluation

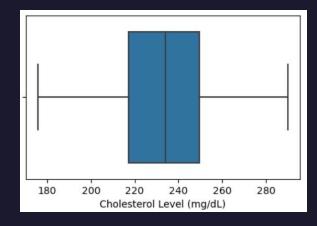




Visualizations to explore the Data







Research Findings

| | OLD. | Regression | Results | | | | |
|-----------------------------|-------------|---------------------------------|-----------|----------|----------|---------|---------|
| Dep. Variable: | Ag | ge (years) | R- | squared | d: (|).928 | |
| Model: | | OLS | Adj. R- | squared | d: (|).928 | |
| Method: | Leas | t Squares | F- | statisti | c: 6 | 416. | |
| Date: | Wed, 22 | Jan 2025 | Prob (F-s | tatistic |): | 0.00 | |
| Time: | | 20:23:27 | Log-Lik | elihood | d: -93 | 884.0 | |
| No. Observations: | | 3000 | | AIC | 1.878 | e+04 | |
| Df Residuals: | | 2993 | | ВІС | 1.882 | +04 | |
| Df Model: | | 6 | | | | | |
| Covariance Type: | n | onrobust | | | | | |
| | | | | | | | |
| | | coe | ef std er | | t P> t | [0.025 | 0.975] |
| | con | st 73.018 | 6 1.620 | 45.08 | 34 0.000 | 69.843 | 76.194 |
| Cholesterol Level (mg/dL) | | L) 0.033 | 5 0.005 | 6.8 | 76 0.000 | 0.024 | 0.043 |
| Blood Glucose Level (mg/dL) | | L) 0.042 | 3 0.006 | 6.78 | 35 0.000 | 0.030 | 0.054 |
| Bone Dens | sity (g/cm | ²) -25.873 | 5 0.454 | -56.9 | 55 0.000 | -26.764 | -24.983 |
| Smoking St | atus_Nev | er -0.573 | 1 0.217 | -2.64 | 46 0.008 | -0.998 | -0.148 |
| Vision Sharpness | | ss -32.614 | 2 0.929 | -35.09 | 0.000 | -34.436 | -30.792 |
| Hearing A | Ability (dl | 3) 0.150 | 5 0.010 | 15.43 | 0.000 | 0.131 | 0.170 |
| | | | | 2.034 | | | |
| Omnibus: | 0.335 | Ourbin-Wat | | | | | |
| Omnibus: | | Ourbin-Wat raue-Bera | | | | | |
| Prob(Omnibus): | | Ourbin-Wat rque-Bera Prob | (JB): | 0.384 | | | |



This model can be used to predict a person's age, thus helping to determine if they are a longevity risk

Formula:

Age (years) = 73.02 + 0.03 (Cholesterol Level) + 0.04 (Blood Glucose Level) - 25.87 (Bone Density)

- 32.61 (Vision Sharpness) + 0.15 (Hearing Ability) - 0.57 (Smoking Status)

Limitations

- Multiple Linear Regression Limitations sensitive to outliers. Outliers must be handled carefully before performing analysis.
- Data Limitations The major life insurance company's data may not include all the relevant health and lifestyle variables.

Next Steps

- Gather comprehensive health and lifestyle information on current customers and their beneficiaries.
- Collect similar health and lifestyle information for new customers and their beneficiaries.



Conclusion

By using a predictive multiple linear regression model, insights can be gained into whether an individual may be considered a longevity risk.

Identifying individuals who are likely to live longer than expected allows life insurance companies to take proactive measures to mitigate unexpected cash flow issues.

