

Modern Statistics:

A Computer-Based Approach with Python

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Modern Statistics

A Computer-Based Approach with Python

Authors: Ron S. Kenett, Shelemyahu Zacks, Peter Gedeck

Demonstrates how to incorporate Python into the modern statistics curriculum

Includes over 40 case studies to facilitate experiential learning

An accompanying Python package is available for download, allowing students to engage directly with the material

Part of the book series: Statistics for Industry, Technology, and Engineering (SITE)

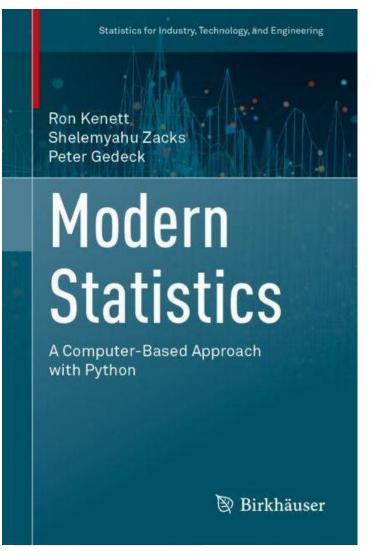


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<u>Keywords</u>

Authors and Affiliations

About the authors

Bibliographic Information



https://gedeck.github.io/mistat-code-solutions/ModernStatistics

- notebooks: Python code of individual chapters in Jupyter notebooks download all as notebooks.zip
- code: Python code for solutions as plain Python files download all as code.zip
- solutions manual : Solutions_Modernstatistics.pdf: solutions of exercises
- solutions: Python code for solutions in Jupyter notebooks download all as solutions.zip
- all : zip file with all files combined download all as all.zip

All the Python applications referred to in this book are contained in a package called mistat available for installation from the Python package index https://pypi.org/project/mistat/. The mistat packages is maintained in a GitHub repository at https://github.com/gedeck/mistat.

Chapter 1: Analyzing Variability: Descriptive Statistics

Chapter 2: Probability Models and Distribution Functions

Chapter 3: Statistical Inference and Bootstrapping

Chapter 4: Variability in Several Dimensions and Regression Models

Chapter 5: Sampling for Estimation of Finite Population Quantities

Chapter 6: Time Series Analysis and Prediction

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Code repository

Installation instructions

Introductory videos



https://gedeck.github.io/mistat-code-solutions/IndustrialStatistics/



Industrial Statistics: A Computer Based Approach with Python is a companion volume to the book Modern Statistics: A Computer Based Approach with Python.

Table of contents (with sample excerpts from chapters)

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Chapter 2: Basic Tools and Principles of Process Control (sample 2)

Chapter 3: Advanced Methods of Statistical Process Control (sample 3)

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Chapter 5: Classical Design and Analysis of Experiments (sample 5)

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Companion book



Analyzing Variability: Descriptive Statistics

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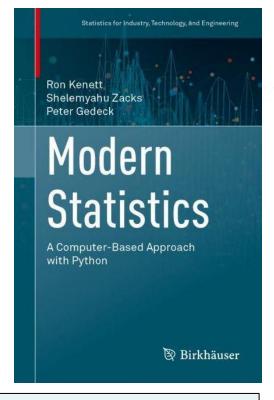
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Modern Statistics: A Computer Based Approach with Python

S

Solutions

Import required modules and define required functions

```
import math
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import mistat
from scipy import stats

def trim_std(data, alpha):
    """ Calculate trimmed standard deviation """
    data = np.array(data)
    data.sort()
    n = len(data)
    low = int(n + alpha) + 1
    high = int(n + (1 - alpha))
    return data[low:(high + 1)].std()
```

Solution 1.1 random.choices selects k values from the list using sampling with replacement.

```
import random
random.seed(1)
values = random.choices([1, 2, 3, 4, 5, 6], k=50)
```

Counter counts the number of occurrences of a given value in a list.

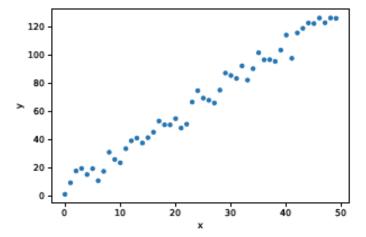
```
from collections import Counter
Counter(values)
```

The expected frequency in each cell, under randomness is 50/6 = 8.3. You will get different numerical results, due to randomness.

Solution 1.2 The Python function range is an iterator. As we need a list of values, we need to explicitly convert it.

```
x = 1ist(range(50))
y = [5 + 2.5 + xi for xi in x]
y = [yi + random.uniform(-10, 10) for yi in y]
pd.DataFrame({'x': x, 'y': y}).plot.scatter(x='x', y='y')
plt.show()
```

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Solution 1.3 In Python

```
from scipy.stats import binom
np.random.seed(1)

for p in (0.1, 0.3, 0.7, 0.9):
   X = binom.rvs(1, p, sixe=50)
   print(p, sum(X))
```

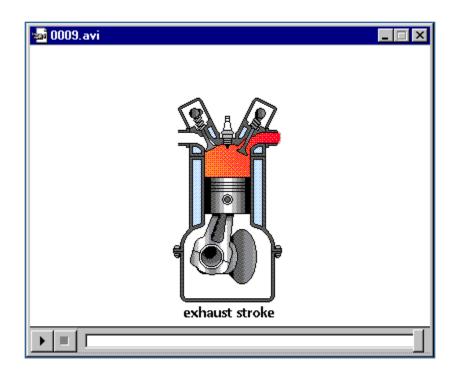
```
0.1 4
0.3 12
0.7 33
0.9 43
```

Notice that the expected values of the sums are 5, 15, 35 and 45.

Solution 1.4 We can plot the data and calculate mean and standard deviation.



The Piston Simulator



D: Spring Coefficient (N/m)

E: Atmospheric Pressure (N/m²)

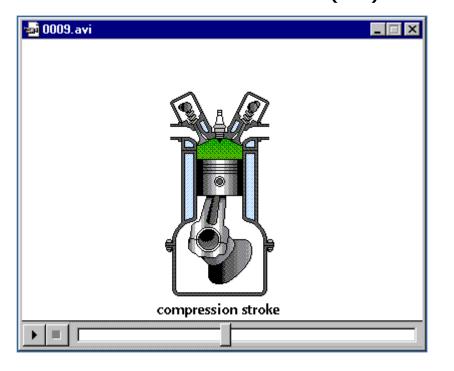
F: Ambient Temperature (⁰K)

G: Gas Temperature (⁰K)

A: Piston Weight (Kg)

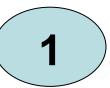
B: Piston Surface Area (m²)

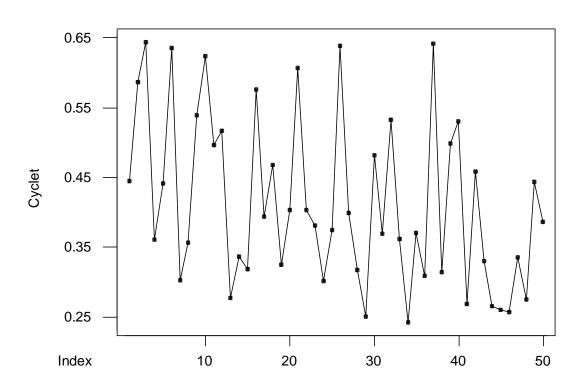
C: Initial Gas Volume (m³)





The Piston Simulator





A run chart of 50 cycle times, in seconds, with all factors set at their maximum value. The cycle times in this sequence have an average of **0.413** seconds and a standard deviation of **0.121** seconds



1.1 Random Phenomena and the Structure of Observations

Many phenomena which we encounter are only partially predictable. It is difficult to predict the weather or the behavior of the stock market. In this book we focus on industrial phenomena, like performance measurements from a product which is being manufactured, or the sales volume in a specified period of a given product model. Such phenomena are characterized by the fact that measurements performed on them are often not constant but reveal a certain degree of variability. Variability is also a reflection of uncertainty. For a comprehensive treatment of uncertainty in engineering applications see del Rosario and Iaccarino (2022). The objectives of this chapter is to present methods for analyzing this variability, in order to understand the variability structure and enhance our ability to control, improve and predict future behavior of such phenomena. We start with a few simple examples.

Example 1.1 A piston is a mechanical device that is present in most types of engines. One measure of the performance of a piston is the time it takes to complete one cycle. We call this measure **cycle time**. In Table 1.1 we present 50 cycle times of a piston operating under fixed operating conditions (a sample data set is stored in file **CYCLT.csv**). We provide with this book code in Python for running a piston software simulation. If you installed Python, install the mistat package using the Python package installer pip.



You will get access to a piston simulator with 7 factors that you can change. We will also use this simulator when we discuss Statistical Process Control (Chapters 2, 3 and 4 in the Industrial Statistics book) and the Design of Experiments (Chapters 5, 6 and 7 in the Industrial Statistics book). We continue at a pedestrian pace by recreating Table 1.1 using Python. All the Python applications referred to in this book are contained in a package called mistat available from the Python package index and on GitHub at https://github.com/gedeck/mistat. The following Python commands will import the mistat package, read the cycle time data and print the first five values on your monitor:

```
import mistat
data = mistat.load_data('CYCLT')
print(data.head())

0    1.008
1    1.098
2    1.120
3    0.423
4    1.021
Name: CYCLT, dtype: float64
```

Notice that functions in Python have parenthesis. The import statement imports the mistat package and makes its functionality available. mistat.load_data is a function that loads the CYCLT data set as a *Pandas* data series. A *Pandas* data series is a simple vector of values.





Example 3.26 Let us determine (0.95,0.95)-tolerance interval for samples of size n = 100, of piston cycle times. Use the sample in the data file **CYCLT.csv**. The original sample is of size $n_0 = 50$. Since future samples are of size n = 100, we draw from the original sample RSWR of size n = 100.

In Python, the calculations from Example 3.26 are straightforward:

```
0.025% 0.175
0.975% 1.141
```

The bootstrap (.95,.95)-tolerance interval for n = 100 piston cycle times was estimated as (0.175,1.141).



Bootstrapping



4

M-test

1. Compute the expected number of returns in category i,

$$E_i = Np_i$$

$$i = 1,...K.$$

2. Compute

$$S_i = SQRT [Np_i(1-p_i)],$$

$$i = 1,...K$$
.

3. Compute the adjusted residuals

$$Z_i = (n_i - E_i)/S_i$$

$$i = 1,...K$$
.

4. Determine the critical value C of Z_i from the M-Test table.

N = Total number of returns

K = Number of categories

n_i = Number of returns in category i

p_i = proportion of category i in sampling frame



M-test

K	10%	5%	1%
4	1.95	2.24	2.81
5	2.05	2.32	2.88
6	2.12	2.39	2.93
7	2.18	2.44	2.99
8	2.23	2.49	3.04
9	2.28	2.53	3.07
10	2.32	2.57	3.10
20	2.57	2.81	3.30
30	2.71	2.94	3.46

If all adjusted residuals, Z_i , are smaller, in absolute value, than C, no significant bias is declared.

Cells with values of Z_i , above C or below -C are declared significantly different and a follow up effort is initiated.

Bonferroni correction



Carlo Bonferroni 1892 - 1960

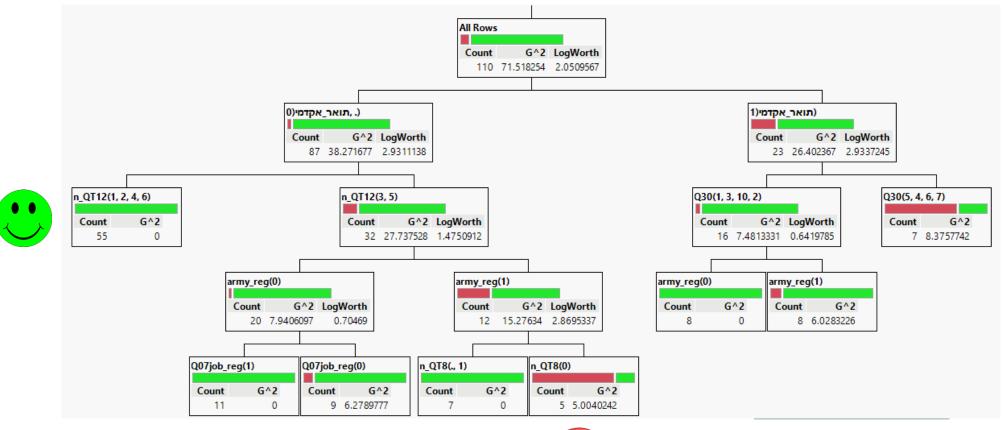


[&]quot;A Test for Detecting Outlying Cells in the Multinomial Distribution and Two-Way Contingency Tables", Fuchs and Kenett, *Journal of the American Statistical Association*, 75, pp. 395-398, 1980.

[&]quot;Two Methods for Comparing Pareto Charts", Kenett, Journal of Quality Technology, 23, pp. 27-31, 1991.

7

Decision trees









ANACONDA, NAVIGATOR









Community

A full Python IDE directly

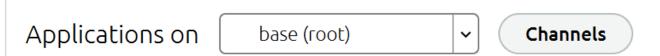
DOCUMENTATION

Δυαζουσα Κιοσ











JupyterLab

3.3.2

An extensible environment for interactive and reproducible computing, based on the Jupyter Notebook and Architecture.





Notebook

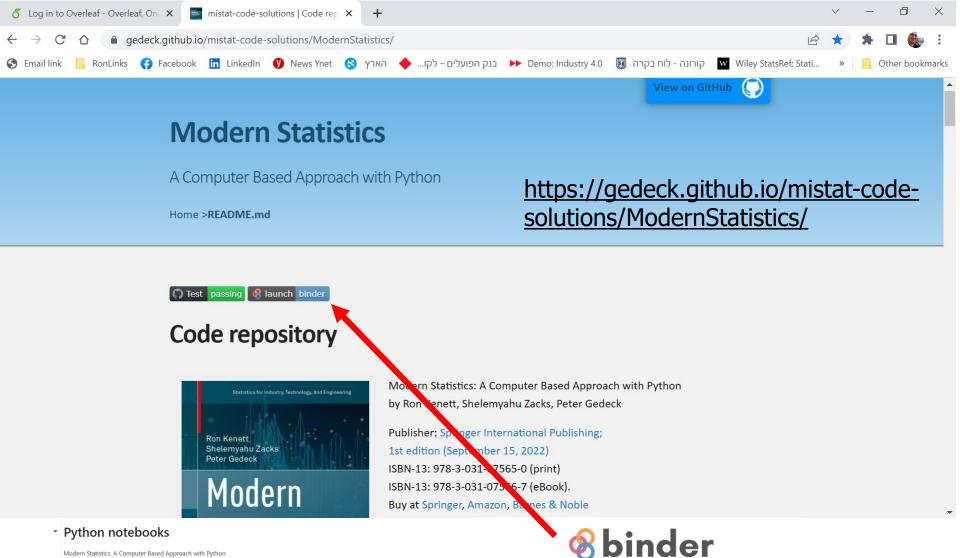
7 6.4.8

Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis.









by Ron Kenett, Shelemyahu Zacks, Peter Gedeck

Publisher: Springer International Publishing; 1st edition (September 15, 2022) ISBN-13: 978-3031075650

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The code needs to be executed in sequence

Python packages and Python itself change over time. This can cause warnings or errors. We "Warnings" are f execution and need to be fixed in order to get results.

If you come across an issue with the code, please follow these steps

- . Check the repository (https://gedeck.github.io/mistat-code-solutions/) to see if the code has been upgr
- . Check the errata at (https://gedeck.github.io/mistat-code-solutions/ModernStatistics/) and the issue tra to see if the issue is known and a solution available.
- Report the problem using the issue tracker at https://github.com/gedeck/mistat-code-solutions/issues

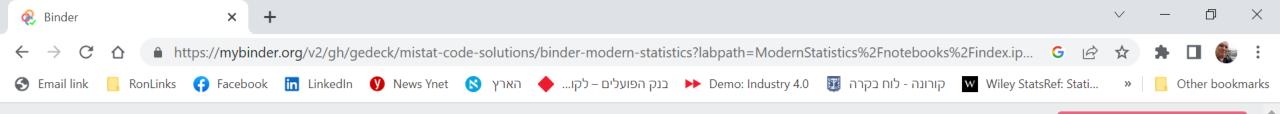




Starting repository: gedeck/mistat-code-solutions/bindermodern-statistics

> Your session is taking longer than usual to start! Check the log messages below to see what is happening.





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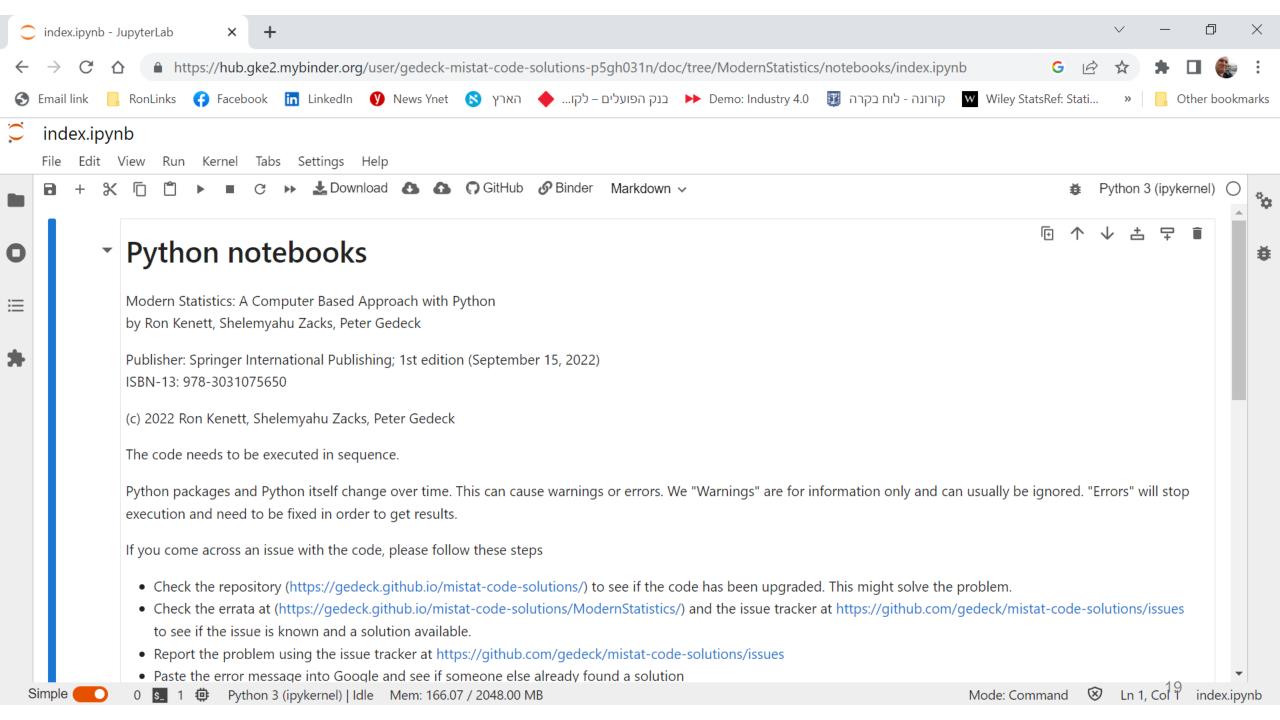


https://mybinder.org/v2/gh/gedeck /mistat-code-solutions/bindermodernstatistics?labpath=ModernStatistics %2Fnotebooks%2Findex.ipynb



Starting repository: gedeck/mistat-code-solutions/bindermodern-statistics

The tool that powers this page is called BinderHub. It is an open source tool that you can deploy yourself.



Modern Statistics

מוסד שמואל נאמן ליומה מדינות לאומת

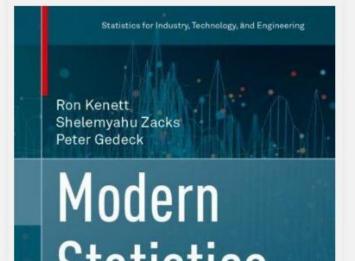
A Computer Based Approach with Python

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https://gedeck.github.io/mistat-code-solutions/ModernStatistics/

Code repository



Modern Statistics: A Computer Based Approach with Python by Ron Kenett, Shelemyahu Zacks, Peter Gedeck

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Modern Statistics: A Computer Based Approach with Python is a companion volume to the book Industrial Statistics: A Computer Based Approach with Python.

This part of the repository contains:

- notebooks : Python code of individual chapters in Jupyter notebooks download all as notebooks.zip
- code: Python code for solutions as plain Python files download all as code.zip
- solutions manual: Solutions Modernstatistics.pdf: solutions of exercises
- solutions: Python code for solutions in Jupyter notebooks download all as solutions.zip
- all: zip file with all files combined download all as all.zip
- datafiles: zip file with all data files download all as data_files.zip the mistat package gives you already access
 to all datafiles, you only need to download this file if you want to use it with different software

All the Python applications referred to in this book are contained in a package called mistat available for installation from the Python package index https://pypi.org/project/mistat/. The mistat packages is maintained in a GitHub repository at https://github.com/gedeck/mistat.

Try the code

You can explore the code on Binder 8 launch binder

Installation instructions

Instructions on installing Python and required packages are here.

These Python packages are used in the code examples of *Modern Statistics*:

- mistat (for access to data sets and additional functionality)
- numpy
- scipy
- scikit-learn



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Chapter 6: Time Series Analysis and Prediction

Chapter 7: Modern analytic methods: Part I

Chapter 8: Modern analytic methods: Part II

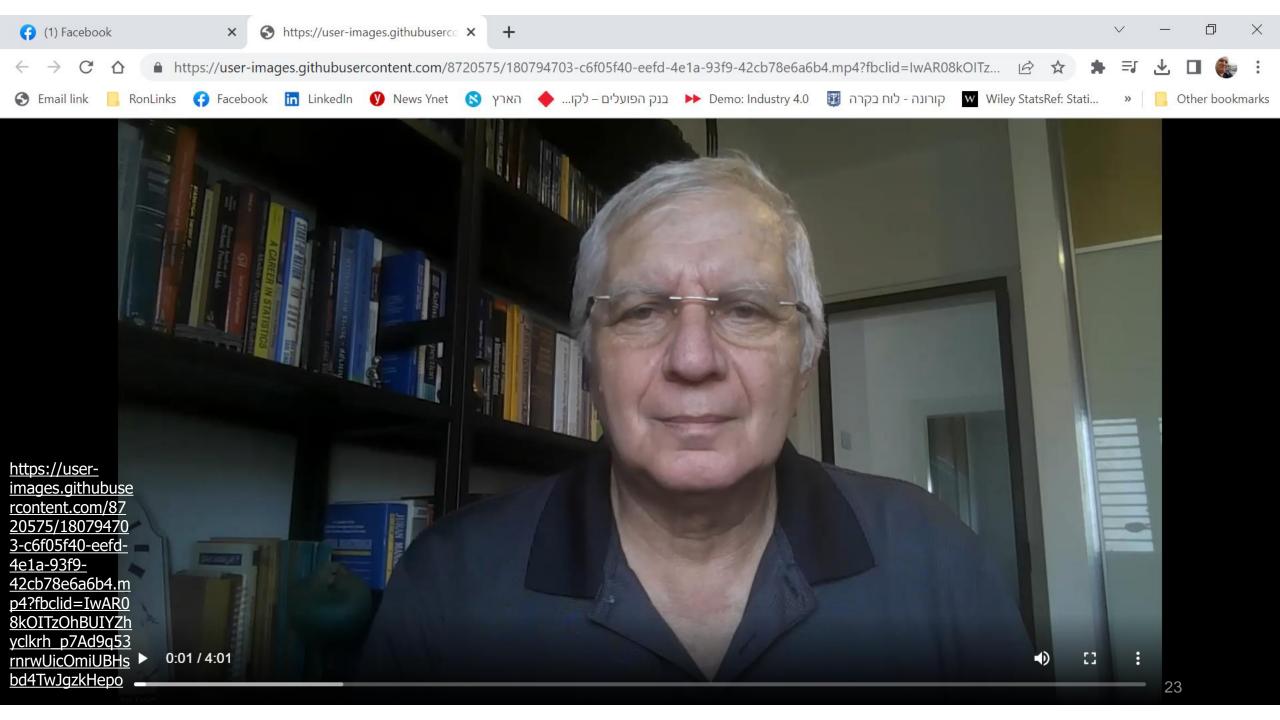
Introductory videos

Chapter 1: Analyzing Variability: Descriptive Statistics

The chapter focuses on statistical variability and on various methods of analyzing random data. Random results of experiments are illustrated with distinction between deterministic and random components of variability. The difference between accuracy and precision is explained. Frequency distributions are defined to represent random phenomena. Various characteristics of location and dispersion of frequency distributions are defined. The elements of exploratory data analysis are presented.











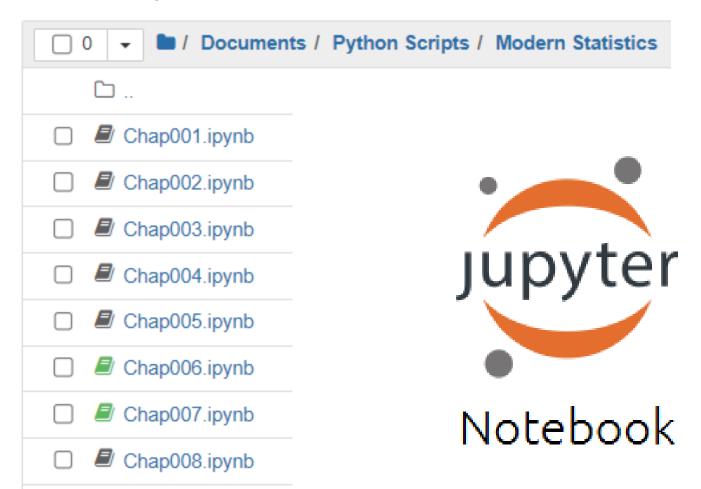
Files

Running

Clusters

Select items to perform actions on them.







Chapter 6

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The code needs to be executed in sequence.

Time Series Analysis and Prediction

Ron Kenett, Shelemyahu Zacks, Peter Gedeck Pages 329-360

```
In [1]:
    import os
    os.environ['OUTDATED_IGNORE'] = '1'
    import warnings
    from outdated import OutdatedPackageWarning
    warnings.filterwarnings('ignore', category=FutureWarning)
    warnings.filterwarnings('ignore', category=OutdatedPackageWarning)
```

Time Series Analysis and Prediction

```
In [2]: import datetime
   import statsmodels.formula.api as smf
   from statsmodels.tools.sm_exceptions import ValueWarning
   import pandas as pd

import random
   import numpy as np
   import pingouin as pg
   from scipy import stats
   import matplotlib.pyplot as plt
   import mistat
```



Chapter 7

Modern Statistics: A Computer Based Approach with Python by Ron Kenett, Shelemyahu Zacks, Peter Gedeck

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In [1]: import warnings
import os
    os.environ['OUTDATED_IGNORE'] = '1'
    from outdated import OutdatedPackageWarning
    warnings.filterwarnings('ignore', category=FutureWarning)
    warnings.filterwarnings('ignore', category=OutdatedPackageWarning)
```

Modern analytic methods: Part I

```
In [2]: import warnings
import random
import pandas as pd
import numpy as np
from sklearn.ensemble import RandomForestClassifier
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score
from sklearn.impute import SimpleImputer
from sklearn.neural_network import MLPClassifier
from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay
```



Modern Analytic Methods: Part I

Ron Kenett, Shelemyahu Zacks, Peter Gedeck Pages 361-393



Chapter 8

Modern Statistics: A Computer Based Approach with Python by Ron Kenett, Shelemyahu Zacks, Peter Gedeck

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The code needs to be executed in sequence.

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In [1]:
    import os
    os.environ['OUTDATED_IGNORE'] = '1'
    import warnings
    from outdated import OutdatedPackageWarning
    warnings.filterwarnings('ignore', category=FutureWarning)
    warnings.filterwarnings('ignore', category=OutdatedPackageWarning)
```

Modern analytic methods: Part II

```
import networkx as nx
import statsmodels.api as sm
from statsmodels.tsa.stattools import grangercausalitytests
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import mistat
```

Modern Analytic Methods: Part II

Ron Kenett, Shelemyahu Zacks, Peter Gedeck Pages 395-419



A Biomed Data Analyst Training Program

Modern Statistics: A Computer Based Approach with Python

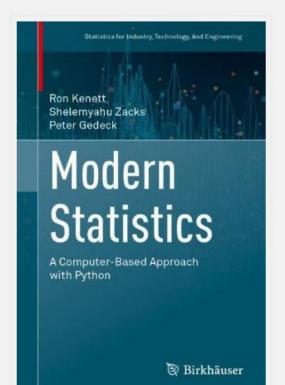
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https://gedeck.github.io/mistat-codesolutions/BioMed DataAnalyst Course/

A Biomed Data Analyst Training Program



Modern Statistics: A Computer Based Approach with Python

by Ron Kenett, Shelemyahu Zacks, Peter Gedeck

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Errata: See known errata here



Slides

וסד שמואל נאמן יוסד מיומר לאומים

- 1. Introduction
- 2. Data types and data integration
- 3. Supervised learning
- 4. Model performance
- 5. Time series
- 6. Data visualization
- 7. Causality and experimental design

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Code and data files

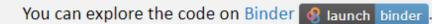
This part of the repository contains:

notebooks : Python code of individual chapters in Jupyter notebooks - download all as notebooks.zip

The Python package mistat contains datafiles and utility functions referred to in the Modern Statistics book. It is available for installation from the Python package index https://pypi.org/project/mistat/. The mistat packages is maintained in a GitHub repository at https://github.com/gedeck/mistat.

Try the code







Thank you for listening

