



Portofolio - Short Class

Data Science Introduction

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Course Summary

Silahkan dapat menuliskan summary dari kelas yang diikuti hari ini.

Poin Belajar	Rangkuman
Introduction to Data Science	Data Science adalah bidang yang menggabungkan statistik, pemrograman, dan analisis data untuk menemukan wawasan yang berharga. Proses dalam Data Science mencakup pengumpulan, pembersihan, eksplorasi, analisis, dan visualisasi data. Data Scientist menggunakan berbagai teknik seperti machine learning dan algoritma statistik untuk membuat prediksi atau keputusan berbasis data. Exploratory Data Analysis (EDA) menjadi tahap penting dalam memahami pola dan distribusi data sebelum membangun model. Selain itu, machine learning memungkinkan sistem untuk belajar dari data tanpa pemrograman eksplisit. Model prediktif digunakan dalam berbagai bidang seperti keuangan, kesehatan, dan pemasaran untuk mengoptimalkan strategi bisnis. Data Science juga memanfaatkan big data dan cloud computing untuk mengelola dan memproses data dalam skala besar. Pemahaman tentang data cleaning sangat penting untuk memastikan kualitas data yang baik sebelum dianalisis. Visualisasi data membantu menyampaikan temuan secara lebih jelas kepada pengambil keputusan. Secara keseluruhan, Data Science berperan penting dalam mengubah data mentah menjadi informasi yang bernilai bagi berbagai industri.



Course Summary

Silahkan dapat menuliskan summary dari kelas yang diikuti hari ini.

Poin Belajar	Rangkuman
The Workflow of Data Science	The Workflow of Data Science mencakup beberapa tahapan utama dalam mengolah data untuk mendapatkan wawasan berharga. Tahap pertama adalah pengumpulan data , yang bisa berasal dari berbagai sumber seperti database, API, atau sensor. Selanjutnya, dilakukan pembersihan data (data cleaning) untuk menangani data yang hilang, duplikat, atau tidak valid. Setelah itu, tahap Exploratory Data Analysis (EDA) dilakukan untuk memahami pola, distribusi, dan hubungan antar variabel. Kemudian, pemrosesan fitur (feature engineering) dilakukan untuk memilih atau membuat variabel yang lebih relevan bagi model. Tahap berikutnya adalah pemilihan dan pelatihan model machine learning dengan algoritma yang sesuai dengan tujuan analisis. Setelah model dilatih, dilakukan evaluasi model menggunakan metrik seperti akurasi, precision, dan recall. Jika hasilnya belum optimal, dilakukan penyetelan parameter (hyperparameter tuning) untuk meningkatkan performa model. Setelah model siap, tahap deploy model dilakukan agar bisa digunakan dalam lingkungan produksi. Terakhir, model yang sudah diterapkan harus dimonitor dan diperbarui secara berkala untuk menjaga akurasinya sesuai dengan perubahan data baru.



```
[94] # import library
  import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
  from google.colab import drive
```



df	<pre>[95] # read and load dataset df = pd.read_csv('/content/drive/MyDrive/Course/Datasets/titanic-case.csv') df.head()</pre>											
∑	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Cabin	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	s
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	s
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	s
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	s



	<pre>[78] # rename variable df = df.rename(columns={'PassengerId': 'passenger_id', 'Survived': 'survived', 'Pclass': 'p_class', 'Name': 'name',</pre>													
∑		passenger_	id surviv	ed p_c	class	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked
	0		1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	NaN	S
	1		2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	C85	С
	2		3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	NaN	s
	3		4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	C123	s
	4		5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	NaN	s



[79] # summary statistics dataset df.describe()												
		passenger_id	survived	p_class	age	sibsp	parch	fare				
	count	891.000000	891.000000	891.000000	714.000000	891.000000	891.000000	891.000000				
	mean	446.000000	0.383838	2.308642	29.699118	0.523008	0.381594	32.204208				
	std	257.353842	0.486592	0.836071	14.526497	1.102743	0.806057	49.693429				
	min	1.000000	0.000000	1.000000	0.420000	0.000000	0.000000	0.000000				
	25%	223.500000	0.000000	2.000000	20.125000	0.000000	0.000000	7.910400				
	50%	446.000000	0.000000	3.000000	28.000000	0.000000	0.000000	14.454200				
	75%	668.500000	1.000000	3.000000	38.000000	1.000000	0.000000	31.000000				
	max	891.000000	1.000000	3.000000	80.000000	8.000000	6.000000	512.329200				







```
[81] # handling missing values in 'age' by filling with the median age
    df['age'].fillna(df['age'].median(), inplace=True)
    # handling missing values in 'embarked' by filling with the mode
    df['embarked'].fillna(df['embarked'].mode()[0], inplace=True)
    # handling missing values in 'fare' by filling with the median fare
    df['fare'].fillna(df['fare'].median(), inplace=True)
    # dropping the 'cabin' column due to a high number of missing values
    df.drop(columns=['cabin'], inplace=True)
```



```
[82] # checking for missing values again
     missing values after = df.isnull().sum()
     print("\nMissing values in the dataset after handling:")
     print(missing values after)
    Missing values in the dataset after handling:
    passenger id
    survived
    p class
     name
     sex
    age
    sibsp
    parch
    ticket
    fare
    embarked
    dtype: int64
```



```
# checking for duplicates
duplicates = df.duplicated().sum()
print("\nNumber of duplicates in the dataset:")
print(duplicates)
Number of duplicates in the dataset:
```



```
[84] # dropping duplicates
     df.drop duplicates(inplace=True)
[85] # checking for duplicates again
     duplicates after = df.duplicated().sum()
     print("\nNumber of duplicates in the dataset after removing duplicates:")
     print(duplicates after)
     Number of duplicates in the dataset after removing duplicates:
```



```
[86] # displaying the first few rows dataset
     print("\nFirst few rows of the dataset after handling missing values and duplicates:")
     print(df.head())
     First few rows of the dataset after handling missing values and duplicates:
        passenger id survived p class \
                                                                       sibsp \
                                 Braund, Mr. Owen Harris
                                                            male
                                                                 22.0
     0
       Cumings, Mrs. John Bradley (Florence Briggs Th...
                                                         female
                                  Heikkinen, Miss. Laina female
                                                                 26.0
            Futrelle, Mrs. Jacques Heath (Lily May Peel)
                                                          female 35.0
                                Allen, Mr. William Henry
                                                            male 35.0
                                                                           0
                                   fare embarked
        parch
                        ticket
                     A/5 21171
                                 7.2500
                      PC 17599 71.2833
              STON/02. 3101282
                                7.9250
                        113803 53.1000
                                 8.0500
                        373450
```



```
[87] # summary statistics dataset
     print("\nSummary statistics of the dataset:")
     print(df.describe())
₹
     Summary statistics of the dataset:
            passenger id
                             survived
                                          p class
                                                                     sibsp \
                                                           age
              891.000000
                          891.000000
                                       891.000000
                                                    891.000000
                                                                891.000000
     count
              446.000000
                             0.383838
                                         2.308642
                                                     29.361582
                                                                  0.523008
     mean
                             0.486592
                                                                  1.102743
     std
              257.353842
                                         0.836071
                                                    13.019697
                1.000000
                             0.000000
                                         1.000000
                                                                  0.000000
     min
                                                     0.420000
                             0.000000
                                         2.000000
                                                    22.000000
                                                                  0.000000
     25%
              223.500000
              446.000000
                             0.000000
                                                    28.000000
     50%
                                         3.000000
                                                                  0.000000
              668.500000
                             1.000000
                                         3.000000
                                                    35.000000
                                                                  1.000000
     75%
     max
              891.000000
                             1.000000
                                         3.000000
                                                    80.000000
                                                                  8.000000
                 parch
                               fare
            891.000000
                        891.000000
     mean
              0.381594
                          32.204208
              0.806057
                          49.693429
     std
     min
              0.000000
                          0.000000
              0.000000
     25%
                          7.910400
     50%
              0.000000
                         14.454200
     75%
              0.000000
                         31.000000
              6.000000 512.329200
     max
```

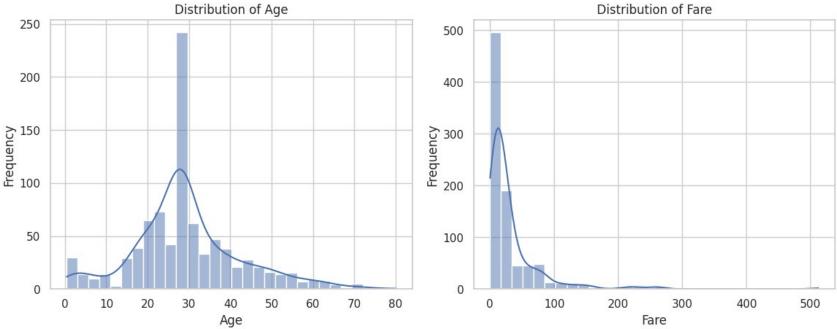


```
[88] # information about the dataset
     print("\nInformation about the dataset:")
     print(df.info())
₹
    Information about the dataset:
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 891 entries, 0 to 890
     Data columns (total 11 columns):
                       Non-Null Count Dtype
         Column
         passenger id 891 non-null
                                       int64
         survived
                       891 non-null
                                        int64
          p class
                       891 non-null
                                        int64
                       891 non-null
                                       object
                                       object
                       891 non-null
                       891 non-null
                                        float64
          sibsp
                                        int64
                       891 non-null
          parch
                       891 non-null
                                        int64
         ticket
                       891 non-null
                                       object
                       891 non-null
                                       float64
        embarked
                       891 non-null
                                       object
     dtypes: float64(2), int64(5), object(4)
     memory usage: 76.7+ KB
     None
```



```
[89] # visualizing the distribution of age and fare
     sns.histplot(df['age'].dropna(), bins=30, kde=True)
     plt.title('Distribution of Age')
     plt.xlabel('Age')
     plt.ylabel('Frequency')
     plt.show()
     sns.histplot(df['fare'].dropna(), bins=30, kde=True)
     plt.title('Distribution of Fare')
     plt.xlabel('Fare')
     plt.ylabel('Frequency')
     plt.show()
```

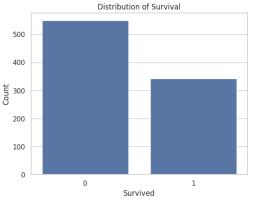


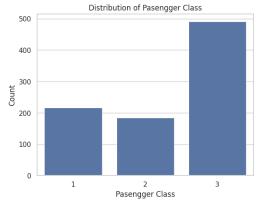


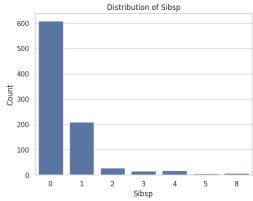


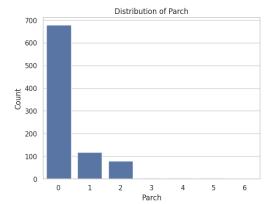
```
[90] # visualizing the distribution of survived, p class, sibsp and parch
     sns.countplot(x='survived', data=df)
     plt.title('Distribution of Survival')
     plt.xlabel('Survived')
     plt.ylabel('Count')
     plt.show()
     sns.countplot(x='p_class', data=df)
     plt.title('Distribution of Pasengger Class')
     plt.xlabel('Pasengger Class')
     plt.ylabel('Count')
     plt.show()
     sns.countplot(x='sibsp', data=df)
     plt.title('Distribution of Sibsp')
     plt.xlabel('Sibsp')
     plt.ylabel('Count')
     plt.show()
     sns.countplot(x='parch', data=df)
     plt.title('Distribution of Parch')
     plt.xlabel('Parch')
     plt.ylabel('Count')
     plt.show()
     sns.countplot(x='embarked', data=df)
     plt.title('Distribution of Embarked')
     plt.xlabel('Embarked')
     plt.ylabel('Count')
     plt.show()
```

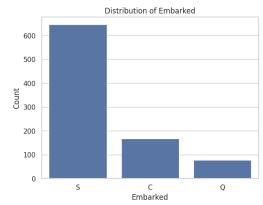










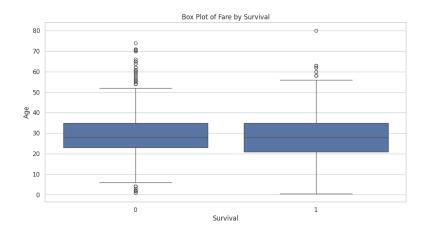


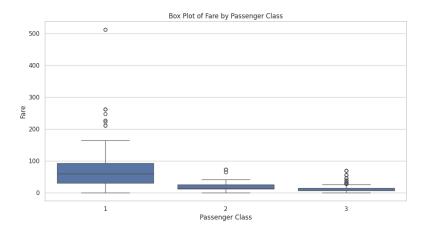
#RintisKarirImpian



```
[91] # box plot for numerical features
     plt.figure(figsize=(12, 6))
     sns.boxplot(x='survived', y='age', data=df)
     plt.title('Box Plot of Fare by Survival')
     plt.xlabel('Survival')
     plt.ylabel('Age')
     plt.show()
     plt.figure(figsize=(12, 6))
     sns.boxplot(x='p class', y='fare', data=df)
     plt.title('Box Plot of Fare by Passenger Class')
     plt.xlabel('Passenger Class')
     plt.ylabel('Fare')
     plt.show()
```







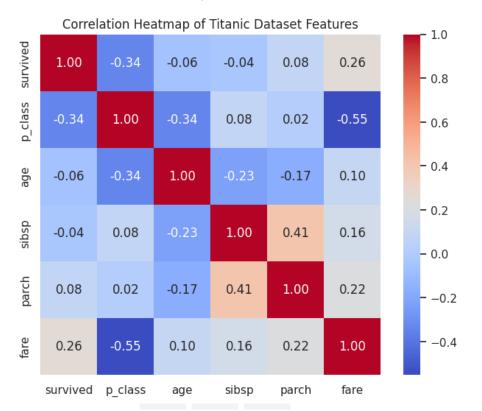


```
# correlation plot heatmap
# Selecting numerical features for correlation analysis
numerical_features = ['survived', 'p_class', 'age', 'sibsp', 'parch', 'fare']
correlation_matrix = df[numerical_features].corr()
# generating the heatmap
plt.figure(figsize=(8, 6))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f', square=True)
plt.title('Correlation Heatmap of Titanic Dataset Features')
plt.show()
```



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Practice using Google Colab







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