

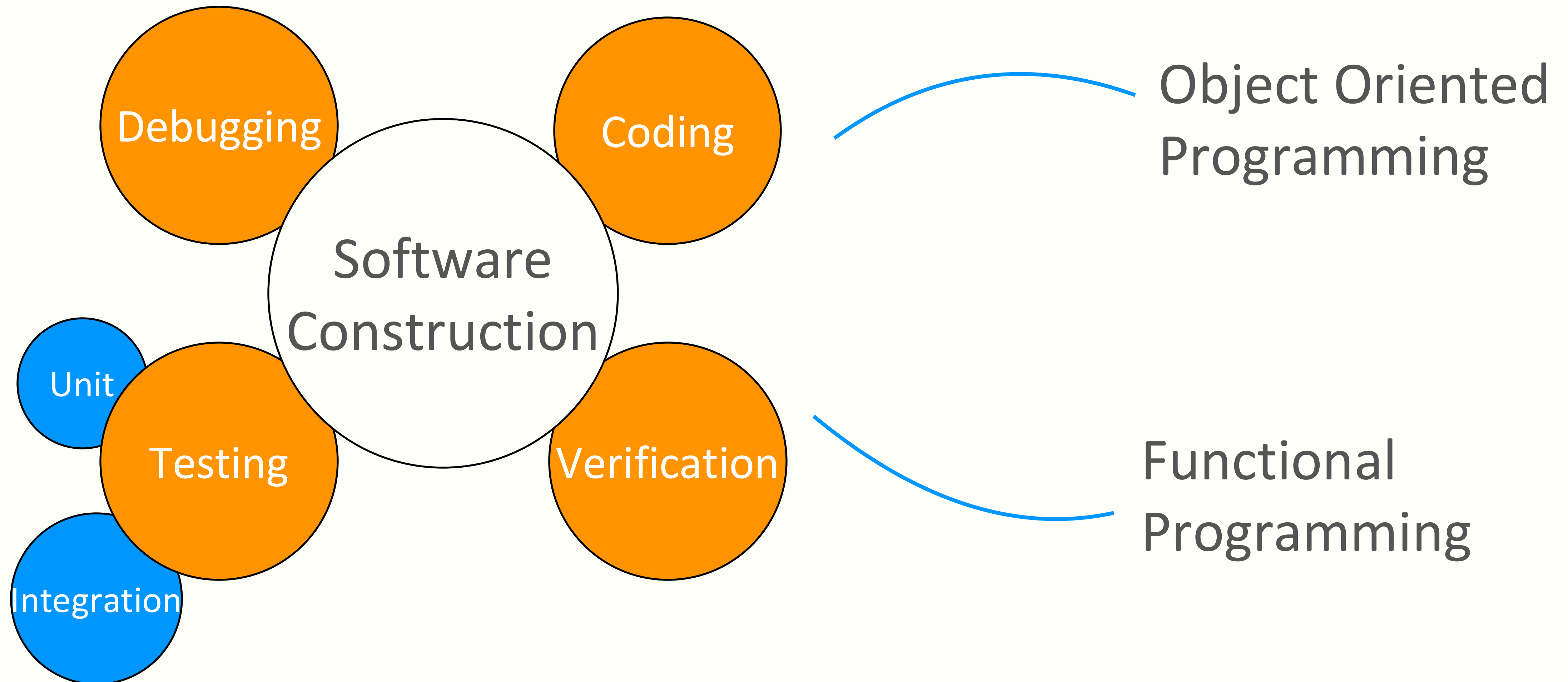
# Object Oriented Programming

# Functional Programming

Perancangan Aplikasi Sains Data  
(Week 5 – 6)

PROGRAM STUDI SAINS DATA  
TELKOM UNIVERSITY

# Introduction to OOP and FP



# Brief History of OOP and FP

Tracing the Development of Programming Paradigms

**Origins of OOP:** Developed in the 1960s with Simula; popularized by Smalltalk, C++, and Java

**Origins of FP:** Rooted in lambda calculus (1930s); adopted in Lisp, Haskell, and modern Python.

**Modern Trends:** Hybrid approaches integrate OOP and FP (e.g., Python, Scala, and JavaScript)

# Understanding The Paradigm

OOP vs. FP: Understanding the Distinctions

**OOP; State & Objects:** OOP focuses on objects that store state and behavior, making it ideal for modular applications

**FP; Pure Functions & Immutability:** FP emphasizes stateless functions and immutable data, enabling parallelism and predictability

**Common Misconceptions:** OOP is not always slower than FP, and FP can be used for complex applications, not just small scripts

# Object Oriented in Data Science

Building Reusable and Scalable Code

**Core OOP concepts:** Encapsulation, inheritance, polymorphism

OOP in Data Science Enhances **code modularity**, **reusability**, and **maintainability** in machine learning pipelines.

Example: Scikit-learn; Scikit-learn models use OOP principles (e.g., fit, predict).

# Functional in Data Science

A Declarative Approach to Data Processing

**Core FP concepts:** Referential Transperency, immutability, higher-order functions.

Facilitates **parallelism**, **reduces side effects**, and **simplifies complex transformations**.

Example: Pandas & PySpark ; Data manipulation using `map()`, `filter()`, `apply()` for scalable processing.

# Functional

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## Principles

Immutability

Referential  
Transparency

Higher Order  
Functions



# Functional

## Immutability

In functional programming, everything is immutable. This means that **once a variable is set, its value cannot be changed**. If `x=3` at the beginning of a program, x will always have the value of 3 for the rest of the program.

### So...

How do we get any work done if we can't change any variable?

```
x = -5  
y = x + 1
```

This code have immutable concept .  
Why?. Give another example!



# Functional

## Referential Transparency

In functional programming, referentially transparent if **we can replace it with its value anywhere in the code**. If you can replace the call of a function with its actual value, then the function is referentially transparent

```
def add(a,b):  
    return a + b  
Print(add(2,3))
```

This code have referential transparency concept . Why?. Give another example!

# Functional

## Higher Order Function

In Functional Programming, functions are first-class citizens, meaning **they can be passed as arguments**, returned from other functions, and stored in variables. A higher order function is a function that takes another function as an argument or returns a function. Higher order functions also **allow us to minimise duplicate code**.

```
def apply_twice(func, x):  
    return func(func(x))  
  
def increment(num):  
    return num + 1  
  
print(apply_twice(increment, 3))
```

This code have referential Higher Order Function concept . Why?. Give the results and another example!

# Functional Task

Analyze this code and explain the concept of functional programming we discussed before (Immutability, referential transparency, and higher-order functions)!

```
def increment_list(numbers):  
    return [num + 1 for num in numbers]  
  
def square(x):  
    return x * x  
  
def apply_function(func, data):  
    return [func(x) for x in data]  
  
# Sample data  
numbers = [1, 2, 3, 4, 5]  
  
# Apply functions  
incremented_numbers = increment_list(numbers)  
squared_numbers = apply_function(square, numbers)  
  
# Print results  
print("Original:", numbers)  
print("Incremented:", incremented_numbers)  
print("Squared:", squared_numbers)
```

Make the source code that demonstrates Immutability, Referential Transparency and Higher Order Function in single running program.

Home  
Work

# End Of Week 5

# Object-Oriented

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## Principles

Inheritance

Encapsulation

polymorphism

# Object Oriented

## Inheritance

Inheritance means that you can **extend a class by creating another class** that builds on it. This helps reduce repetition, because if you need a new class that's closely related to one you have already written, you don't need to duplicate that class to make a minor change.

## So...

Explain the inheritance method in that code!. Give another example!

```
# Parent class (Base class)
class Hewan:
    def __init__(self, name):
        self.name = name

    def bunyi(self):
        return "bunyi hewan"

# Child class
class Bebek(Hewan):
    def bunyi(self):
        return "Kwek-kwek!"

# Child class
class Tikus(Hewan):
    def bunyi(self):
        return "Gludug-Gludug!"

Bebek = Bebek("Donald")
Tikus = Tikus("Mickey")

# Same method from parent, different result
# depend on clid
print(Bebek.name, "Bunyi:", Bebek.bunyi())
print(Tikus.name, "Bunyi:", Tikus.bunyi())
```



# Object Oriented

## Encapsulation

In Encapsulation means that **your class hides its details from the outside**. You can see only the interface to the class, not the internal details of what's going on. The interface is made up of the methods and attributes that you design. **It's not so common in Python**, but in other programming languages classes are often designed with hidden or private methods or attributes that can't be changed from the outside.

**So...**

Which gives error? ❶ or ❷ . Why?

```
class BankAccount:
    def __init__(self, saldo):
        self.__saldo = saldo

    def deposit(self, jumlah):
        """Increase balance"""
        if jumlah > 0:
            self.__saldo += jumlah
        return self.__saldo

    def cek_saldo(self):
        """Return the current balance"""
        return self.__saldo

account = BankAccount(1000)
account.deposit(500)
#print("Saldo:", account.cek_saldo) ❶
#print(account.__saldo) ❷
```

# Object Oriented

## Polymorphism

Polymorphism means that you can have the **same interface for different classes**, which simplifies your code and reduces repetition. That is, two classes can have a method with the **same name that produces a similar result, but the internal workings are different**. The two classes can be a parent and child class, or they can be unrelated.

**So...**

Where the polymorphism concept in that code?, and Add several shapes with polymorphism concept!

```
class Bentuk:
    def luas(self):
        return 0

class Persegi(Bentuk):
    def __init__(self, sisi):
        self.sisi = sisi

    def luas(self):
        return self.sisi ** 2

class Lingkaran(Bentuk):
    def __init__(self, jejari):
        self.jejari = jejari

    def luas(self):
        return 3.14 * self.jejari ** 2

bentuk = [Persegi(4), Lingkaran(3)]

for bentuk in bentuk:
    print("Luas:", bentuk.luas())
```

# Object Oriented

## Simple Example

```
class DataProcessor:
    def __init__(self, data):
        self.data = data # Store data inside the object

    def clean_data(self):
        """Remove negative values from data"""
        self.data = [x for x in self.data if x >= 0]
        return self.data # Return processed data

# Example usage
processor = DataProcessor([10, -5, 20, -2, 30])
print(processor.clean_data()) # Output: [10, 20, 30].
```

## Why OOP?

- Ideal Groups related functions (clean\_data) **inside a class**.
- Encapsulation: The data is stored **inside the DataProcessor** object.

# Functional

## Simple Example

```
data = [10, -5, 20, -2, 30]

# Functional transformations
clean_data = list(filter(lambda x: x >= 0, data)) # Remove
negatives
squared_data = list(map(lambda x: x ** 2, clean_data)) #
Square the values

print(squared_data) # Output: [100, 400, 900]
```

### Why FP?

- Uses **pure functions** (**map()**, **filter()**) without modifying the original list.
- No side effects, making it **more predictable and reusable**.

# Object Oriented - Functional

Choosing the Right Paradigm



## OOP Strengths

Best for structured, reusable, and large-scale applications (e.g., machine learning pipelines).



## FP Strengths

Ideal for data transformations, parallel computing, and immutable workflows.



## Hybrid

Many real-world applications mix both paradigms for flexibility and performance.



# Why it's Matter?

## The Role of OOP & FP in Data Science

**Code Organization:** Programming paradigms define how code is structured, improving readability and maintainability.

**Performance Optimization:** Programming paradigms define how code is structured, improving readability and maintainability.

**Scalability in Data Science:** OOP and FP enable scalable solutions for data science, from preprocessing to machine learning.

# When We Use It?

Choosing the Right Paradigm for Your Use Case

**Use OOP When:** Building large, structured applications like ML model deployment & API'S.

**Use FP When:** Processing large datasets with transformations, parallel computing, and immutability.

**Hybrid Approach:** Many data science projects combine OOP for structure and FP for transformations.



# Case Study

```
import pandas as pd
from IPython.display import display

class CekHargaBeras:
    def __init__(self, df):
        self.df = df.copy()
        # Mengisi missing value
        self.df['Harga_Beras'].fillna(self.df['Harga_Beras'].mean(), inplace=True)

    def get_data(self):
        return self.df

# Contoh Data
df = pd.DataFrame({
    'Provinsi': ['Jakarta', 'Jawa Barat', 'Jawa Tengah', 'Bali'],
    'Harga_Beras': [12000, None, 11000, 11500]
})

# Process Data
processor = CekHargaBeras(df)
processed_data = processor.get_data()

# Display
display(processed_data)
```

# Case Study

Identify, why this source code adopt OOP concept?. Make the program in FP!

```
from sklearn.linear_model import LinearRegression

class ModelWrapper:
    def __init__(self):
        self.model = LinearRegression()

    def train(self, X, y):
        """Train the model on the given data"""
        self.model.fit(X, y)
        return self

    def predict(self, X):
        """Make predictions using the trained model"""
        return self.model.predict(X)

# Example usage
X = processed_data[['A']] # Feature
y = [5, 10, 15, 20] # Target variable

# Train model
model = ModelWrapper().train(X, y)

# Make predictions
predictions = model.predict(X)
print(predictions)
```

# Home Work

Identify and analyze this code.  
What concept of OOP/FP used in  
that code and where?

```
class MachineLearningPipeline:
    def __init__(self, df, target_column):
        self.df = df
        self.target_column = target_column
        self.processor = DataProcessor(df)
        self.model = ModelWrapper()

    def run(self):
        """Complete ML workflow: preprocess data, train model, predict"""
        processed_df = self.processor.clean_data().scale_features(['A']).get_data()
        X, y = processed_df[['A']], self.df[self.target_column]
        self.model.train(X, y)
        return self.model.predict(X)

# Example usage
df['Target'] = [5, 10, 15, 20] # Add target column
pipeline = MachineLearningPipeline(df, 'Target')
predictions = pipeline.run()
print("Final Predictions:", predictions)
```

# Home Work

Identify and analyze this code.  
What concept of OOP/FP used in  
that code and where?

# End Of Week 6