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Higher Order Functions

Currying, Collections and Callbacks, and Abstract Data
Types

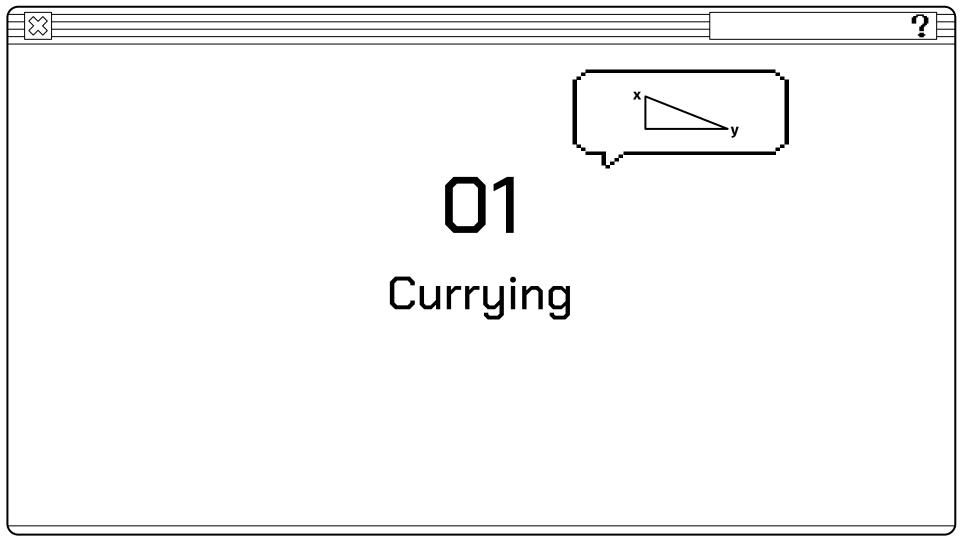
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O1 Currying How and why it is used?

O2 Collections and Callbacks

Map, FlatMap, FoldRight, and Other Folds and Reduce

O3 Abstract Data Types Map, Set, Parallel and Distributed



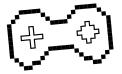


Currying

Definition: The process of transforming a function with multiple variables into multiple functions, each taking a single argument.

Transformation:

- A function f(x, y) that takes two arguments can be transformed into f(x)(y).
- The first function takes x and returns a new function that takes y.



Example



def add(x: Int, y: Int): Int = x + y // Regular function
with two arguments

def addCurried(x: Int)(y: Int): Int = x + y // Curried

How it works:

version

- Call addCurried(3) → This returns a function
 (y: Int) ⇒ 3 + y.
- Call that returned function with 4 → addCurried(3)(4) results in 7.



Benefits

Peartial Applications

Allows fixing some arguments and reusing the resulting function.

val addThree = addCurried(3) // Partially
applied function
println(addThree(4)) // Outputs: 7

Key Concept

Currying turns a function into a chain of functions, each handling one argument at a time.



Exercise

Suppose you are tasked with implementing a curried function in Scala that performs the following:

- 1. Takes a String prefix as its first argument.
- 2. Takes an Int multiplier as its second argument.
- 3. Takes a list of numbers as its third argument.
- 4. Returns a new list where each number is multiplied by the given multiplier, converted to a String, and prefixed with the given prefix.

show an example where you fix the prefix and multiplier using partial application, then process multiple lists of numbers.

```
O1 Define the Curried Function:
def processList(prefix: String)(multiplier: Int)(numbers: List[Int]): List[String] = {
```

```
Partial Application:
   val prefixWithMultiplier = processList("Result: ")(2) // Fix prefix as "Result: " and
   multiplier as 2
```

```
val list1 = List(1, 2, 3)
val list2 = List(4, 5, 6)

O4 Apply the partially applied function to lists:
```

val result1 = prefixWithMultiplier(list1)

Define lists of numbers:

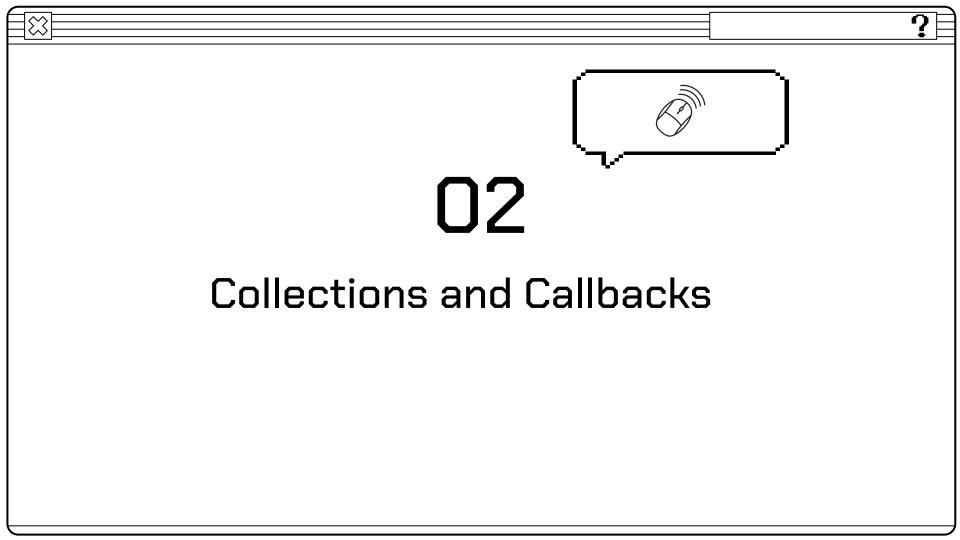
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numbers.map($n \Rightarrow prefix + (n * multiplier).toString)$

```
val result2 = prefixWithMultiplier(list2)

O5 Print:
```

```
println(result1) // Output: List("Result: 2", "Result: 4", "Result: 6")
println(result2) // Output: List("Result: 8", "Result: 10", "Result: 12")
```





Maps



Purpose: Transforms each element of a collection into a new element using a provided callback function.

Signature: def map[B](f: A \Rightarrow B): List[B]

- Takes a function f that maps each element A to B.
- Returns a new collection containing the transformed elements.

Example:

val numbers = List(1, 2, 3) val doubled = numbers.map($x \Rightarrow x * 2$) // List(2, 4, 6) println(doubled)



flatMap

Purpose: Similar to map, but the callback function returns a collection rather than a single element. flatMap flattens the resulting collections into a single collection.

Signature:

def flatMap[B](f: A \Rightarrow List[B]): List[B]

- <u>Use Case:</u> When each element can be mapped to zero or more elements.

Example:

val numbers = List(1, 2, 3) val expanded = numbers.flatMap($x \Rightarrow List(x, x * 10)$) println(expanded) // List(1, 10, 2, 20, 3, 30)

 $x \Rightarrow List(x, x * 10)$ maps each number to a list of two elements, and flatMap combines all these lists into a single list.





foldRight

Purpose: Combines elements of a collection using a binary operation, starting from the rightmost element.

Signature:

def foldRight[B](z: B)(op: (A, B) \Rightarrow B): B

- z is the starting value (accumulator). op is a binary operation: it combines each element with the current accumulator.
- Right-to-Left Processing: The operation starts from the end of the list and moves to the beginning.

Example:

val numbers = List(1, 2, 3)val sum = numbers.foldRight(0)((x, acc) \Rightarrow x + acc) // 1 +(2+(3+0))println(sum) // Output: 6





foldLeft

Purpose: Similar to foldRight, but processes elements left-to-right (from the start of the list).

Signature:

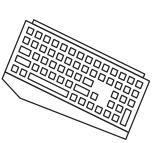
def foldLeft[B](z: B)(op: (B, A) \Rightarrow B): B

Example:

val numbers = List(1, 2, 3) val sum = numbers.foldLeft(0)((acc, x) \Rightarrow acc + x) // ((0 + 1) + 2) + 3 println(sum) // Output: 6

Key Difference:

- foldLeft: Left-to-right, accumulator is updated
 first.
- foldRight: Right-to-left, last element is processed first.







Purpose: Combines all elements of a collection using a binary operation, without an initial value.

Signature: def reduce(op: $(A, A) \Rightarrow A$): A

Key Point: Unlike fold, reduce requires the collection to be non-empty because there is no initial value (z).

Example:

```
val numbers = List(1, 2, 3)
val sum = numbers.reduce((x, y) \Rightarrow x + y) // 1 + 2 + 3
println(sum) // Output: 6
```

- directly.Use flatMap when each element results in a collection.
- Use foldLeft or foldRight for accumulation with an initial value.
- Use reduce for accumulation when no initial value is needed.

- Use map when transforming elements



Exercise

You have a list of students, where each student is represented as a case class: case class Student(name: String, scores: List[Int])

For example:

val students = List(Student("Alice", List(85, 90, 78)), Student("Bob", List(92, 88, 95)), Student("Charlie", List(70, 65, 80)))

Your tasks are:

Task 1: Use map to create a list of students' average scores as a tuple: (name, averageScore).

Task 2: Use flatMap to generate a flat list of all scores.

Task 3: Use foldLeft to calculate the total sum of all scores in the class.

Task 4: Use reduce to find the student with the highest average score.



```
Ol Define the Data and Case Class:
    case class Student(name: String, scores: List[Int])
    val students = List(
        Student("Adrian", List(85, 90, 78)),
        Student("Bob", List(92, 88, 95)),
        Student("Charlie", List(70, 65, 80))
```

```
O2 Use map to Get (Name, Average Score):
   val nameAndAverage = students.map(student ⇒ {
      val average = student.scores.sum.toDouble / student.scores.size
      (student.name, average)
})
```

(Bob, 91.66666666666667), (Charlie, 71.6666666666667))

// Output: List(85, 90, 78, 92, 88, 95, 70, 65, 80)

Use flatMap to Get a Flat List of All Scores: val allScores = students.flatMap(.scores)

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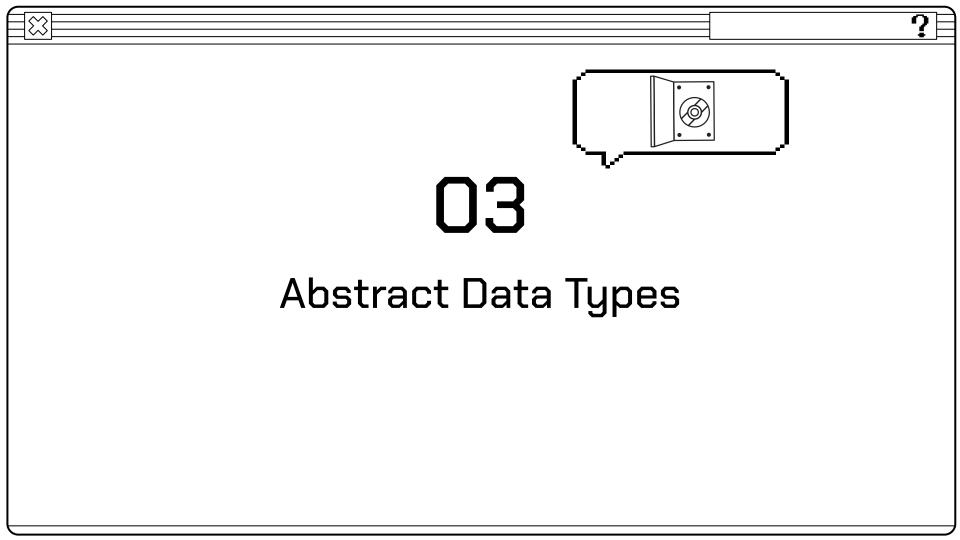
println(allScores)

println(nameAndAverage)// Output: List((Adrian, 84.3333333333333),

Use foldLeft to Calculate the Total Sum of All Scores val totalScore = allScores.foldLeft(0)((acc, score) ⇒ acc + score) println(totalScore) // Output: 743

O5 Use reduce to Find the Student with the Highest Average Score: First, calculate the averages for all students, and then use reduce to compare their averages.

```
val topStudent = nameAndAverage.reduce((student1, student2) ⇒
   if (student1._2 > student2._2) student1 else student2
)
println(topStudent)
// Output: (Bob,91.6666666666667)
```





Sets

Definition: A Set is a collection of unique elements with no duplicates.

Kev Characteristics:

- Commonly used for membership testing, union, intersection, and difference operations.
- Immutable sets are more common in functional programming, ensuring immutability guarantees.

Key Operations:

- set.contains(elem): Checks if the element exists in the set.
- set + elem: Returns a new set with the element added.
- set elem: Returns a new set with the element removed.
- Set Operations: union, intersection, and difference.

Example

```
val set1 = Set(1, 2, 3)
val set2 = Set(3, 4, 5)
```

```
val union = set1.union(set2)
// Set(1, 2, 3, 4, 5)
val intersection =
set1.intersect(set2) //
Set(3)
val difference =
set1.diff(set2) // Set(1, 2)
println(union)
```

println(intersection)

println(difference)

Parallel and Distributed Collections

Definition: Parallel and distributed collections are used to process large datasets efficiently by dividing work across multiple CPU cores or distributed systems.

Key Characteristics:

- Operations like map, reduce, and fold are executed in parallel or distributed across nodes.
- Functional programming provides guarantees like immutability, ensuring that operations can run safely in parallel.

Parallel Collections in Scala:

- ParSeq, ParMap, etc. parallelize operations like map, filter, and reduce.
- Use .par

Distributed Collections:

- In distributed programming, collections are processed across multiple machines (e.g., Apache Spark's RDDs).
- Functions like map, flatMap, and reduce are distributed and executed on different partitions of the dataset.



Exercise

Design a Scala program that processes a large dataset of employees and their work details. The dataset consists of employee records, represented as: case class Employee(id: Int, name: String, department: String, salary: Double, projects: Set[String])

```
Given this list of employees:
```

val employees = List(

```
Employee(1, "Alice", "Engineering", 85000.0, Set("ProjectA", "ProjectB")),
    Employee(2, "Bob", "Engineering", 90000.0, Set("ProjectC")),
    Employee(3, "Charlie", "HR", 60000.0, Set("ProjectD", "ProjectE",
"ProjectF")),
    Employee(4, "Diana", "Engineering", 120000.0, Set("ProjectA")),
    Employee(5, "Eve", "HR", 65000.0, Set("ProjectE", "ProjectF")),
    Employee(6, "Frank", "Marketing", 70000.0, Set("ProjectG"))
)
```



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O1 Task 1: Filter Employees by Department and Salary:

val highEarningEngineers = employees.filter(emp ⇒ emp.department = "Engineering" & emp.salary > 80000)
println(highEarningEngineers)

O2 Task 2: Map to Names and Project Counts:
 val engineerProjectCounts = highEarningEngineers.map(emp ⇒ (emp.name,

emp.projects.size))
println(engineerProjectCounts)
// Output: List((Alice,2), (Bob,1), (Diana,1))

Task 3: Generate a Flat List of All Unique Project Names:

val allProjects = employees.flatMap(_.projects).toSet
println(allProjects)
// Output: Set(ProjectA, ProjectB, ProjectC, ProjectD, ProjectE, ProjectE

// Output: List(Employee(1,Alice,Engineering,85000.0,Set(ProjectA, ProjectB)), ...

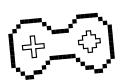
// Output: Set(ProjectA, ProjectB, ProjectC, ProjectD, ProjectE, ProjectF, ProjectG)

O4 Task 4: Calculate Total Salary Using FoldLeft
val totalSalary = employees.foldLeft(0.0)((acc, emp) ⇒ acc + emp.salary)
println(totalSalary)
// Output: 490000.0

O5 Task 5: Find the Employee with the Highest Salary Using Reduce val highestPaidEmployee = employees.reduce((emp1, emp2) ⇒ if (emp1.salary > emp2.salary) emp1 else emp2) println(highestPaidEmployee) // Output: Employee(4,Diana,Engineering,120000.0,Set(ProjectA))

Task 6: Compute Total Salary Using Parallel Collections
val parallelTotalSalary = employees.par.map(_.salary).reduce(_ + _)
println(parallelTotalSalary)
// Output: 490000.0





Thanks!

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Source:https://csci3155.cs.colorado. edu/pppl-course/book/higher-order-fu nctions.html#sec-abstract-data-types -and-higher-order-functions

