

Reactive Programming

J. Heinzelreiter

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 - Distributed Programming with Actors
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Literature (1)

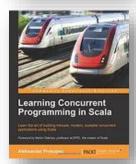
[Odersky, 2021]: Martin Odersky, Lex Spoon, Bill Venners.
 Programming in Scala. Artima, 5th ed., 2021.



 [Pilquist et al., 2023]: Michael Pilquist, Rúnar Bjarnason, and Paul Chiusano. Functional Programming in Scala. Manning, 2nd ed., 2023.



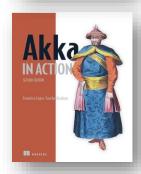
[Prokopec, 2017]: Aleksandar Prokopec. Learning Concurrent
 Programming in Scala. Packt Publishing, 2nd ed., 2017.



Literature (2)

- [Odersky et al., 2015]: Martin Odersky, Erik Meijer, Roland Khun: Principles of Reactive Programming. Online Course, 2015 (see YouTube).
- [Roestenburg et al., 2017]: R. Roestenburg, R. Bakker, R. Williams. Akka in Action. Manning, 2017.
- [Lopez-Sancho Abraham, 2023]: F. Lopez-Sancho Abraham.
 Akka in Action. Manning, 2nd ed., 2023.
- [Kuhn, 2017]: R. Kuhn. Reactive Design Patterns. Manning, February 2017.





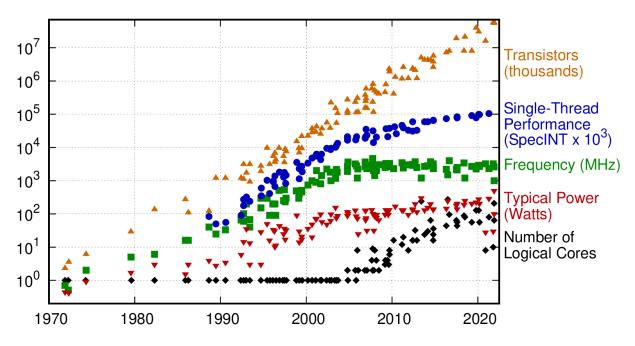


Motivation for Reactive Programming

[Odersky et al., 2015; What is Reactive Programming?] [http://www.reactivemanifesto.org]

Moore's Law

- Moore's law is the observation that the number of transistors on integrated circuits doubles approximately every two years.
- Clock speed doesn't increase correspondingly anymore.
- Consequences:
 - Shift from single- to multi-core computers.
 - Concurrent programming is becoming more and more important.
 - Modern programming languages are becoming functional.



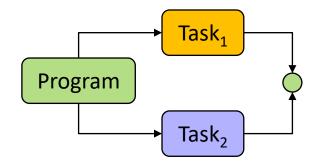
https://github.com/karlrupp/microprocessor-trend-data

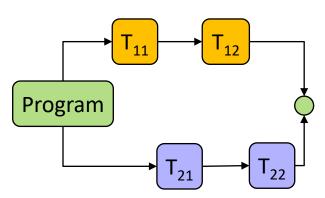
"The Free Lunch Is Over"

- Article from Herb Sutter (2005): The Free Lunch Is Over.
 A Fundamental Turn Toward Concurrency in Software
 - Instead of driving clock speeds and straight-line instruction throughput ever higher, they are instead turning en masse to hyper-threading and multicore architectures.
 - Applications will increasingly need to be concurrent if they want to fully exploit continuing exponential CPU throughput gains.
 - Efficiency and performance optimization will get more, not less, important.
 - The vast majority of programmers today don't grok concurrency, just as the vast majority of programmers 15 years ago didn't yet grok objects.

Concurrent vs. Parallel

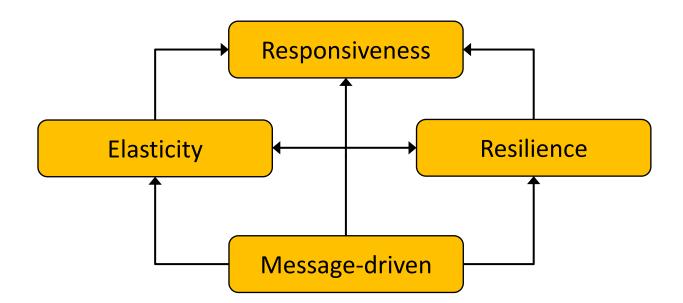
- Parallelism: Techniques to accelerate programs by performing computations simultaneously.
 - Requires multiple CPUs
 - Computation units must be independent
 - Goal: Increase runtime efficiency
- Concurrency: Program makes progress on multiple tasks at the same time.
 - Concurrent programs can be executed on a single CPU, but can benefit from multiple CPUs
 - Goal: Allow efficient interaction with multiple external agents
 - Concurrency includes parallelism





Reactive Systems

- Definition of reactive: "Showing a response to a stimulus. Pupils are reactive to light" [Oxford Dictionaries]
- The Reactive Manifesto: Characteristics of Reactive Systems



Characteristics of Reactive Systems (1)

- Message-driven: react to events
 - Reactive Systems rely on asynchronous message-passing \rightarrow no blocking.
 - Ensures loose coupling of components \rightarrow location transparency.
 - Message queues enable load management.
 - Also, failures are communicated via messages.
- Elasticity: react to load
 - The system stays responsive under varying workload.
 - Scale up: make use of multiple cores, increase memory
 - Scale out: make use of multiple server nodes
 - Requirements:
 - Minimize mutable state
 - Location transparency, resilience

Characteristics of Reactive Systems (2)

- Resilience: react to failures
 - The system stays responsive in the face of failure.
 - Parts of the system can fail and recover without compromising the system as a whole.
 - Recovery of each component is delegated to another component.
 - Prerequisites:
 - Encapsulation of state
 - Supervisor hierarchies
- Responsiveness: react to users
 - System must provide real-time interaction with agents under load and in the presence of failures.
 - Prerequisites: Message-driven architecture, elasticity, resilience.

Traditional Concurrent Programming Techniques

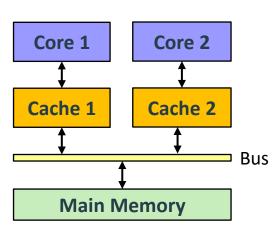
[Prokopec, 2017; p. 27 ff.]

Overview

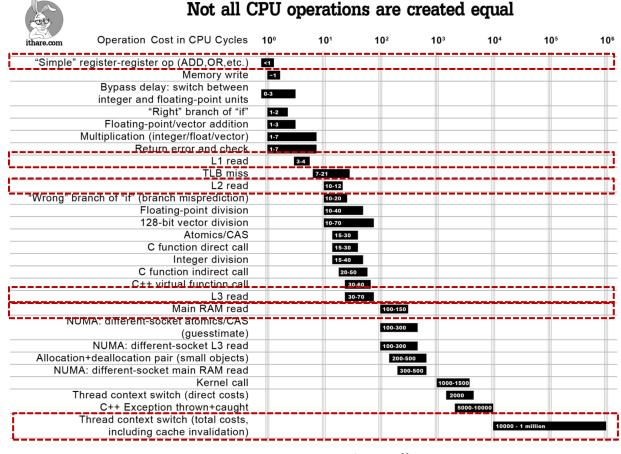
- Traditional low-level constructs for concurrent executions:
 - Processes and
 - Threads
- Means for communication:
 - Shared memory communication → synchronization required
 - Message passing → distributed systems
- Problem: low level and error prone
 - Deadlocks
 - Race conditions
 - Data races
 - Starvation
- Basic knowledge of low-level approaches is essential to understand high-level concepts.

Processes and Threads

- A process executes the instructions of a program.
 - Shares CPU and other resources with other processes.
 - Has its own share of memory.
- A process' instructions are executed concurrently in multiple threads.
 - Threads share the memory of the process they belong to.
 - Threads communicate by writing and reading to memory.
- Threads can be executed on different processors (cores).
 - Processors do not write directly to main memory.
 - They use caches to improve read/write performance.
 - One processor cannot access the cache of another processor.
- Java threads are directly mapped to OS threads.
- Scala inherits the thread model of Java.



Costs of different CPU operations



Distance which light travels while the operation is performed













http://ithare.com/infographicsoperation-costs-in-cpu-clock-cycles

Java Memory Model (JMM)

- Scala inherits the memory model from the JVM.
- The memory model defines when writes to a variable are visible to other threads.
- The compiler and the runtime perform various optimizations to gain performance.
 - Registers may be used as intermediate storage.
 - Data may be written to hierarchies of caches.
 - Bytecode statements may be reordered.
- The rules of the JMM define how threads interact through memory:
 - Program order: Program optimizations by the compiler must not alter the serial semantics within a thread.
 - Locking: Locks for the same synchronization variable must not overlap.
 - Volatile fields: A write to a volatile field is immediately visible to all threads.
 - Thread start: All actions in a thread are done after a call to start().
 - Many other rules.

Working with Threads in Scala

Thread creation

```
class MyThread extends Thread:
    override def run(): Unit =
        println("Executed in new thread.")

val t = new MyThread
t.start()
```

Convenience method to execute code in separate thread

```
def doInThread(body: => Unit): Thread =
  val t = new Thread:
    override def run() = body
  t.start()
  t
```

```
val t = doInThread { println("Executed in new thread.") }
```

Execution of Multi-Threaded Code

If multiple threads access the same memory locations, the program behavior is not deterministic anymore.

- The program works correctly if execution of statements (1) and (2) is not interfered by the other thread.
- t1 and t2 generate the same id if both threads concurrently read the field currld.
- We say that there is a race condition in a program when the output of a program depends on the execution schedule of the instructions.

Synchronization

- Every Java/Scala object has a special property called an intrinsic lock or monitor.
- Only one thread can acquire ownership of a lock.
- If a thread tries to acquire the lock while another thread owns the lock, this thread is blocked.
- If a thread executes the statement x.synchronized { statements }
 - it tries to acquire the lock for x which is released at the end of the block.
 - The thread is blocked while x is locked by another thread.
- The race condition in the example can be removed by locking the region accessed by multiple threads:

```
var currId = 0L

def generateUniqueId() = this.synchronized
  val newId = currId + 1
  currId = newId
  newId
```

Data Races and Reordering

How can this program result in x==1 && y==1?

```
var a = false; var b = false
var x = -1;    var y = -1

val t1 = doInThread { a = true; y = if (b) 0 else 1 }
val t2 = doInThread { b = true; x = if (a) 0 else 1 }

t1.join(); t2.join()
assert(!(x == 1 && y == 1))
```

- It's required that a as well as b are false to produce x==1 && y==1 as a result.
- Possible reasons for this "abnormal" behavior:
 - Reordering of statements: This reordering has no impact on serial semantics but influences behavior of concurrent execution.
 - Data race: Write of one thread (to cache) is not seen be the other thread.
- The synchronized keyword also guarantees that writes to memory of one thread are visible to all other threads.

Deadlocks

Synchronization can result in deadlocks:

```
class Account(val name: String, var balance: Double);
def transfer(amount: Double, a: Account, b: Account): Unit =
    a.synchronized {
    b.synchronized {     a.balance -= amount; b.balance += amount }
}
```

Deadlock can be avoided when resources are always locked in the same order:

```
def transfer(amount: Double, a1: Account, a2: Account): Unit =
  if (a1.name < a2.name)
    a1.synchronized { a2.synchronized { ... } }
  else
    a2.synchronized { a1.synchronized { ... } }</pre>
```

In practice it's difficult to ensure a consistent ordering of resources.

Guarded Blocks

Busy-waiting can be avoided by using Java's wait/notify mechanism.

```
class Consumer(val tasks: Queue[() => Unit]) extends Thread:
    def addTask(task: => Unit) = tasks.synchronized {
        tasks.enqueue(() => task)
        tasks.notifyAll();
    }
    private def getTask(): () => Unit = tasks.synchronized {
        while (tasks.isEmpty) tasks.wait() // guarded block
        tasks.dequeue();
    }
    override def run() =
        while true do
        val task = getTask()
        task()
```

- Wait can cause spurious wakeups: Condition thread is waiting for is not met.
- Condition must be checked repetitively \rightarrow guarded block.

Volatile Variables

- Reads and writes to volatile variables are atomic.
- Operations (eg. incrementation) are not atomic.
- Writes to volatile variables are immediately visible to all threads.
 - Variables are not cached or held in registers.
 - Compiler will not reorder instructions.
- Thread communication via volatile variables is very fast.
- Most common application: status flags

```
class Worker extends Thread:
    @volatile
    private var stopped = false
    override def run() =
        while !stopped do { /* do some work */ }
    def shutdown() = stopped = true
}
```

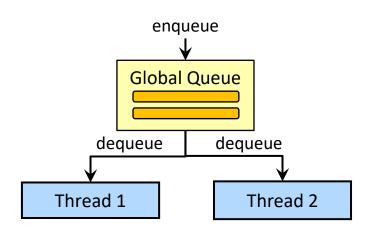
This program may not terminate if @volatile is omitted.

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Thread Pools

- Creating processes is very expensive, creating threads is expensive.
- Thread pools manage a set of threads.

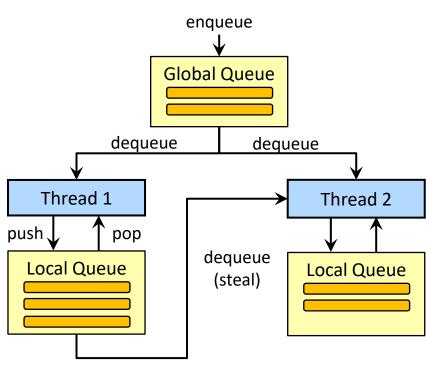
ThreadPoolExecutor



Advantages of ForkJoinPool

- Efficient for recursive tasks
- Less contention in local queue
- Load balancing achieved with work stealing

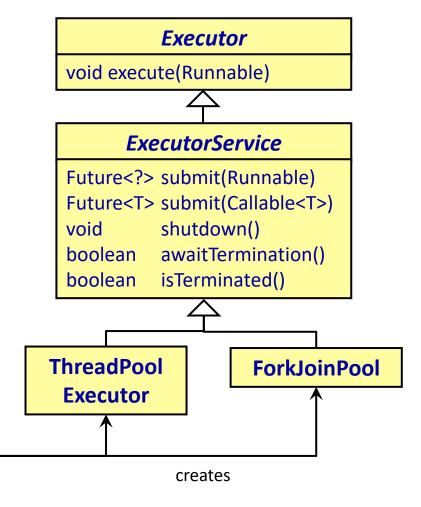
ForkJoinPool



Executor and Executor Service

- Executor decides on which thread and when to execute a task.
- Executor decouples business logic from threading infrastructure.
- Tasks must implement Runnable or Callable<T>.
- Callable objects return values.
- More advanced concepts, like Completable-Futures leverage thread pools internally.
- Scala uses Java's thread pool infrastructure.

Executors (S) newFixedThreadPool() (S) newWorkStealingPool()



Using Executors in Scala

Executor implementations can be used directly in Scala.

```
val executor = new java.util.concurrent.ForkJoinPool
executor.execute(
   () => println("This task is run asynchronously."))
```

- ExecutionContext is similar, but more Scala specific.
- Internally ExecutionContext uses a ForkJoinPool instance.

```
val execCtx = scala.concurrent.ExecutionContext.global
execCtx.execute(
   () => println("This task is run asynchronously."))
```

- Often an ExecutionContext object is passed implicitly to methods.
- A customized ForkJoinPool instance can also be used:

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Atomic Variables

 Atomic variables support complex operations (more than one read/write), that are executed atomically.

```
private val uid = new AtomicLong(0L)
def getUniqueId(): Long = uid.incrementAndGet()
```

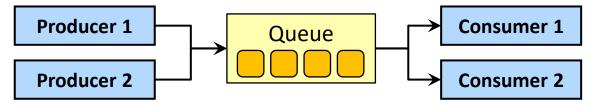
- Operations are implemented lock-free
 - Operations are very fast
 - No danger of deadlocks
- Operations are implemented in terms of a fundamental atomic operation: compareAndSet (also called compare-and-swap).
 - Takes the expected previous value and the new value
 - Fails if previous value changed → operation has to be repeated

```
private val currId = new AtomicLong(0L)
@tailrec def generateUniqueId(): Long =
  val oldId = currId.get
  val newId = oldId + 1
  if currId.compareAndSet(oldId, newId) then newId else generateUniqueId()
```

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Concurrent Collections

- Producer-Consumer is a common pattern in concurrent programming.
 - Work items need to be buffered in concurrent queue



- Java's concurrent queues can also be used in Scala.
- BlockingQueue supports three types of operations, depending on the behavior when queue is empty or full:
 - Operations throw an exception: add/remove/element
 - Operations return a special value: offer/poll/peek
 - Operations block: put/take
- Iterators are weakly consistent: modifications might or might not be reflected in iterator.

Foundations of Functional Programming

[Pilquist et al., 2023]

High-Order Functions

- Functions which take other functions as parameters are called high-order functions.
- It's easy to create and to handle functions: Functions are first-class citizens in Scala.
- Example:

```
def sumSquares(a: Int, b: Int): Int =
  if a > b then 0 else a * a + sumSquares(a + 1, b)

def sumPowerOfTwo(a: Int, b: Int): Int =
  if a > b then 0 else Math.pow(2, a).toInt + sumPowerOfTwo(a + 1, b)
```

```
def sum(f: Int => Int, a: Int, b: Int): Int =
   if (a > b) 0 else f(a) + sum(f, a + 1, b)

def sumSquares(a: Int, b: Int): Int = sum(x => x * x, a, b)

def sumPowerOfTwo(a: Int, b: Int): Int = sum(Math.pow(2, _).toInt, a, b)
```

 Parameterization of functions by other functions is a typical design pattern in functional programming.

Immutable State

Typical procedural Java code:

```
public class Product { ... }
public static List<String> outOfStock(List<Product> allProds) {
   List<Product> prods = new ArrayList<>();
   for (Product p : allProds) if (p.inStock < MIN_STOCK) prods.add(p);
   Collections.sort(prods, new Comparator<Product>() { ... });
   List<String> prodNames = new ArrayList<String>();
   for (String p : prods) prodNames.add(p.getName());
   return prodNames;
}
```

- Functional Scala code is cleaner, because
 - no state has to be tracked,
 - code describes what has to be done and not how it has to be done.

Applying Functions to Collections

filter composes a collection of all elements for which a given predicate holds:

```
val evenNumbers = List(1, 2, 3, 4, 5) filter (_ % 2 == 0) // List(2, 4)
```

map applies a function to each element of a collection:

```
val lowerCase = List("a", "b", "c")
val upperCase = lowerCase map (_.toUpperCase) // List("A", "B", "C")
```

```
val lowerUpper = lowerCase map (str => List(str, str.toUpperCase))
  // List(List("a","A"), (List("b", "B"), List("c", "C"))
```

- flatMap
 - Applies a function that generates a sequence out of each element of a collection.
 - Flattens the resulting sequence of sequences to a (flat) sequence.

```
val lowerUpper = lowerCase flatMap (str => List(str, str.toUpperCase))
  // List("a", "A", "b", "B", "c", "C")
```

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For Expressions (1)

• map and flatMap can be cascaded \rightarrow code may be hard to understand:

- yields in List((Adam, Cain), (Eve, Cain), (Eve, Abel))
- using map would yield in List(List((Adam, Cain)), ..., List())
- The following for expression does exactly the same:

- The compiler translates the second query into the first one.
- The second query is much clearer to read.

For Expressions (2)

Filters

can also be mapped to for expressions

For expressions with side effects

```
for (p <- persons) println(p.name)</pre>
```

are mapped to the function foreach

```
persons.foreach (p => println(p.name));
```

 A for expression can be applied to all domains that provide map, flatMap, filter/ withFilter (and foreach).

Monads

- There are many types that define map and flatMap.
- A type M that supports the three operations flatMap, map, and unit is called a monad:

map can be defined in terms of flatMap and unit:

```
def map[U](f: T \Rightarrow U) : M[U] = m flatMap (x <math>\Rightarrow unit(f(x)))
```

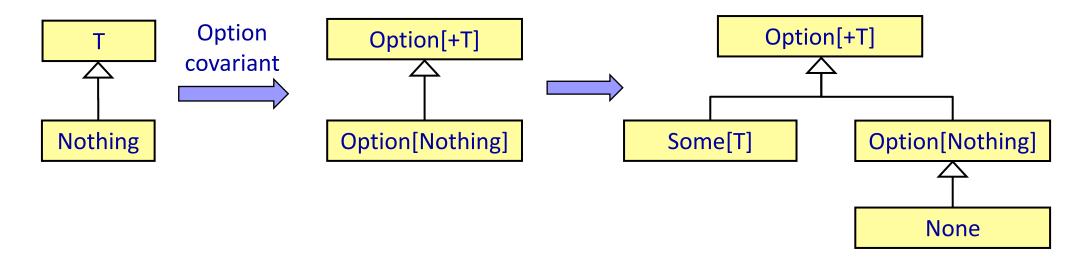
- In a stricter definition monads must fulfill certain laws.
- In other functional languages flatMap is called bind.
- A monad need not be a collection. Often monads handle a single value.
- In Scala the name of the unit function is monad-specific.

The Option Monad

- Options instances hold values of two forms:
 - Some(x) where x is the actual value
 - None represents an undefined value

- Advantages over null
 - Cleaner interfaces: It's obvious that value may be undefined.
 - Checking for undefined value is enforced.

The Option Monad: Embedding into the Type System



Application:

```
val op1: Option[String] = Some("Hello")
val op2: Option[String] = None
```

The Option Monad: map

The unit operations Some an None are generators for monad values:

```
val o1: Option[Double] = Some(9)
Val o2: Option[Double] = None
```

map transforms Option values to other Option values:

```
def toDouble(value: String): Double = value.toDouble

Some("9") map (n => toDouble(n)) // \rightarrow Some(9.0)

None map (n => toDouble(n)) // \rightarrow None
```

map allows for function composition:

```
def sqrt(value: Double): Double = math.sqrt(value)
Some("9") map (n => toDouble(n)) map (n => sqrt(n)) // \rightarrow Some(3.0)
```

Applying a function to a None value does not result in an exception:

```
None map (n \Rightarrow toDouble(n)) map (n \Rightarrow sqrt(n)) // \rightarrow None
```

The Option Monad: flatMap

map is not always useful when applied to option values because it generates "nested objects":

```
def toDouble(value : String) : Option[Double] =
  try { Some(value.toDouble) }
  catch { case nfe : NumberFormatException => None }
  Some("9") map (n => toDouble(n)) // → Some(Some(9.0))
  Some("x") map (n => toDouble(n)) // → Some(None)
```

This can be prevented by applying flatMap:

```
Some("9") flatMap (n => toDouble(n)) // \rightarrow Some(9.0)
Some("x") flatMap (n => toDouble(n)) // \rightarrow None
```

In this manner functions can also be composed:

```
def sqrt(value : Double) : Option[Double] = value match { ... } 

Some("9") flatMap (n => toDouble(n)) flatMap (n => sqrt(n)) // \rightarrow Some(3.0) 

Some("x") flatMap (n => toDouble(n)) flatMap (n => sqrt(n)) // \rightarrow None 

Some("-9") flatMap (n => toDouble(n)) flatMap (n => sqrt(n)) // \rightarrow None
```

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Monad Laws

Associativity

```
(m flatMap f) flatMap g == m flatMap (x => f(x) flatMap g)

(Some("4") flatMap (toDouble(_))) flatMap (sqrt(_)) ==
   Some("4") flatMap (x => toDouble(x) flatMap (sqrt(_))) == Some(2.0)
(None flatMap (toDouble(_))) flatMap (sqrt(_)) ==
   None flatMap (x => toDouble(x) flatMap (sqrt(_))) == None
```

Left unit

```
unit(x) flatMap f == f(x)

Some("1") flatMap (toDouble(_)) == toDouble("1") == Some(1.0)
```

Right unit

```
m flatMap unit == m

Some("1") flatMap (Some(_)) == Some("1")
None flatMap (Some(_)) == None
```

Monads and For Expressions

- Because monads support map and flatMap they can be used in for expressions.
- Function composition expressions like

can also be written as for expressions:

```
val res =
  for (o1 <- Some("4");
      o2 <- toDouble(o1);
      o3 <- sqrt(o2)) yield o3;</pre>
```

Advantage: better readability

The Try Monad (1)

- Try can have two types of values:
 - Success(x): Computation ended successfully with result x.
 - Failure(ex): Computation failed with Exception ex.

- Fatal exceptions: OutOfMemoryError, StackOverflowError, InterruptedException, etc.
- Application:
 - Composition of functions that may fail
 - Pass results between functions, threads and processes

The Try Monad (2)

Like each monad Try must define flatMap and map:

```
abstract class Try[+T]:
    def flatMap[U](f: T => Try[U]): Try[U] = this match
        case Success(x) => try f(x) catch { case NonFatal(ex) => Failure(ex) }
        case fail: Failure => fail

    def map[U](f: T => U): Try[U] = this match
        case Success(x) => Try(f(x)) // Try.apply(f(x))
        case fail: Failure => fail
```

In the stricter interpretation Try is not a monad because the left unit law does not hold:

```
Try(expr) flatMap f != f(expr)
```

 The left-hand side never throws a (non-fatal) exception whereas the right-hand side will raise any exception thrown by f.

The Try Monad: map

The unit operations Success and Failure are generators for monad values:

```
val success: Try[String] = Success("9")
val failure: Try[String] = Failure(IllegalArgumentException())
```

map transforms Try values to other Try values:

```
def toDouble1(value: String): Double = value.toDouble
Success("9") map (n => toDouble1(n)) // → Success(9.0)
Failure(ex) map (n => toDouble1(n)) // → Failure(ex)
```

map allows for function composition:

```
def sqrt1(value: Double): Double = value match
  case v: Double if v >= 0 => math.sqrt(value)
  case _ => throw IllegalArgumentException()
Success("9") map (n => toDouble1(n)) map (n => sqrt1(n)) // → Success(3.0)
```

Applying a function to a Failure value does not result in an exception:

```
Failure(ex) map (n => toDouble1(n)) map (n => sqrt1(n)) // \rightarrow Failure(ex)
```

The Try Monad: flatMap

map is not always useful when applied to Try values because it generates "nested objects":

```
def toDouble2(value: String): Try[Double] = Try { value.toDouble }
Success("9") map (n ⇒ toDouble2(n)) // → Success(Success(9.0))
Success("x") map (n ⇒ toDouble2(n)) // → Success(Failure(NumberFormatException))
```

This can be prevented by applying flatMap:

```
Success("9") flatMap (n => toDouble2(n)) // \rightarrow Success(9.0) Success("x") flatMap (n => toDouble2(n)) // \rightarrow Failure(NumberFormatException)
```

In this manner functions can also be composed:

The Try Monad: For Expressions

Because Try supports map and flatMap, Try can be used in for expressions:

Try objects can be processed by pattern matching:

```
r1 match
  case Success(x) => println(s"Success($x)")
  case Failure(ex) => println(s"Failure($ex)")
```

Futures and Promises

[Prokopec, 2017; p. 101 ff.]

Asynchronous Programming

- In certain situations, threads have to wait for external resources (e. g. data arriving over the network).
- Traditional programs block threads in such cases.
- Possible problems with blocking synchronization:
 - Poor utilization of threads
 - Danger of deadlocks
 - Starvation of threads pools
- Asynchronous programming:
 - Thread has to wait for resource \rightarrow separate computation is scheduled.
 - Thread continues.
 - Original computation is continued when resource is available.
 - Prevents expensive context switches.
- Futures and Promises are abstractions that support asynchronous programming.

Futures

A future

```
trait Future[T]:
    def value: Option[Try[T]]

object Future:
    def apply[T](computation: => T): Future[T]
```

- is a placeholder for a value of type T → future value
- executes an asynchronous computation \rightarrow future computation
- Starting a future computation:

```
val f : Future[Int] = Future {
  computeNthPrime(1000)
} // shorthand for Future.apply(...)
```

Accessing the future value (never do this in this way in production code):

```
println(f.value) // → None
while (! f.isCompleted) { Thread.sleep(100) }
println(f.value) // → Some(Success(7919))
```

ExectionContext

- Future computations are carried out in an ExecutionContext.
- The ExecutionContext is passed as the 2nd parameter to Future.apply:

```
object Future:
   def apply[T](computation :=>T)
        (using executor: ExecutionContext): Future[T]
```

The Scala library provides a default implementation for this implicit parameter:

```
import scala.concurrent.ExecutionContext.Implicits.global
```

- The global ExecutionContext is backed by ForkJoinPool
 - The actual/minimum/maximum number of threads is set to Runtime.availableProcessors.
- The ExectionContext can be easily replaced:

```
given ExecutionContext =
   ExecutionContext.fromExecutor(Executors.newFixedThreadPool(10))
```

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Future callbacks (1)

 Callbacks are (partial) functions that are called when the future computation succeeds or fails.

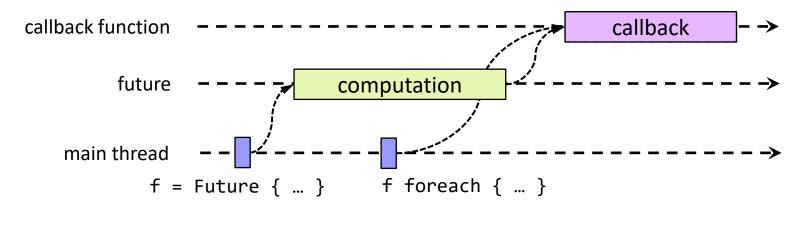
- Futures callbacks are called eventually after the computation is finished and independently of other callbacks.
- Example:

```
val f: Future[List[Int]]= Future { getPrimes(100) }
f foreach { primes => println(primes.mkString(", ")) }
f.failed foreach { ex => println(s"Failed with $ex") }
f onComplete {
  case Success(primes) => println(primes.mkString(", "))
  case Failure(ex) => println(s"Failed with $ex")
}
```

Future callbacks (2)

- Installing a callback is a non-blocking operation (like future creation).
- Callback may or may not be executed in a separate thread.

```
val f = Future { computation }
f foreach { callback }
```



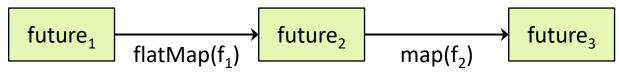
----- happens-before relationship

Functional Composition of Futures (1)

- Problems related to callbacks
 - Cascading futures becomes cumbersome (nested callbacks)
 - Combining the results of multiple futures is difficult
 - Error handling becomes confusing
- Functional composition is a pattern in which functions are cascaded using high-order functions, so called *combinators*.
- Monads provide combinators and are therefore an important means for functional composition.



 Futures support combinators like map, flatMap, filter, etc. that can be used to build pipelines of future computations.



Functional Composition of Futures (2)

- Combinators generate futures out of futures → combinators don't block
- Combinators provided by Future:

- map: Apply function f to future value. f is executed in a new future. Exceptions are propagated to the new future.
- flatMap: Similar to map, but result is not nested

```
val f1: Future[Int] = Future { 1 };
val f2 = f1 map { i => Future { i+1 } } → f2: Future[Future[Int]]
val f1: Future[Int] = Future { 1 };
val f2 = f1 flatMap { i => Future { i+1 } } → f2: Future[Int]
```

• recover: Function pf handles exceptions thrown in future computations. Valid results will be propagated to the new future.

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Functional Composition of Futures (3)

Example for recover:

Example for map:

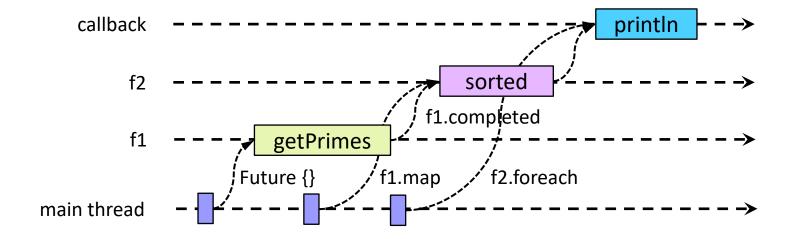
```
val f1: Future[List[Int]] = Future { getPrimes(100) }
val f2: Future[List[Int]] = f1 map { primes => primes.sorted }
f2 foreach { primes => println(primes.mkString(", ")) }
```

Example for flatMap:

```
val f1 = Future { getPrimes(2, 50) }
val f2 = Future { getPrimes(51, 100) }
val f3 = f2 flatMap { primes2 => f1 map { primes1 => primes1 ++ primes2 } }
val f4 = f3 map { primes => primes.sorted }
f4 foreach { primes => println(primes.mkString(", ")) }
```

Semantics of the map Combinator

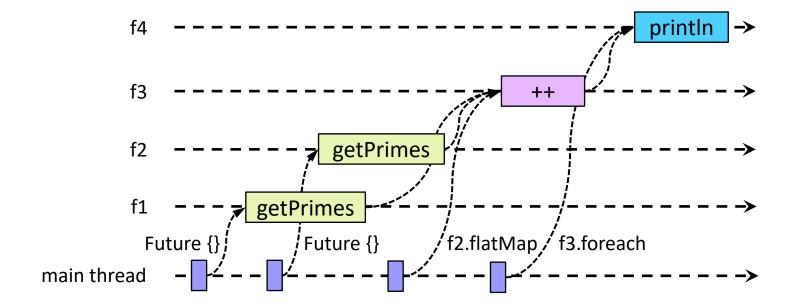
```
val f1: Future[List[Int]] = Future { getPrimes(100) }
val f2: Future[List[Int]] = f1 map { primes => primes.sorted }
f2 foreach { primes => println(primes.mkString(", ")) }
```



Futures and combinators may (but need not) be executed in separate threads.

Semantics of the flatMap Combinator (1)

```
val f1 = Future { getPrimes(2, 50) }
val f2 = Future { getPrimes(51, 100) }
val f3 = f2 flatMap { primes2 => f1 map { primes1 => primes1 ++ primes2 } }
f3 foreach { primes => println(primes.mkString(", ")) }
```



Semantics of the flatMap Combinator (2)

- Difference between Future.flatMap and Future.map:
 - Let f: Future[U], g: U => V
 - val res = f flatMap { u => Future { g(u) } }
 - Execution order:
 - Future f completes with a future value of type U.
 - Future { g(u) } completes with a future value of type V.
 - Future res completes with a future value of type V.
 - Return type: Future[V].
 - val res = f map { u => Future { g(u) } }
 - Execution order:
 - Future f completes,
 - Future res completes with a future value of type Future[V],
 - Future { g(u) } completes.
 - Return type: Future[Future[V]].

Futures and For Expressions

- Since futures support the methods map, flatMap, foreach (and withFilter), futures can be combined in for expressions.
- Therefore, this program fragment

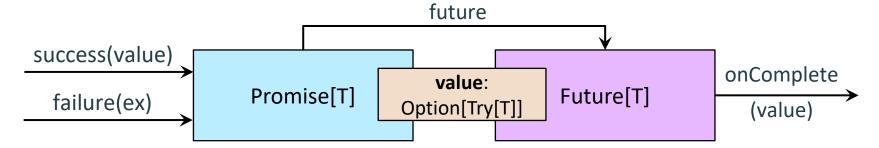
```
val f1 = Future { getPrimes(2, 50) }
val f2 = Future { getPrimes(51, 100) }
val f3 = f2 flatMap { primes2 => f1 map { primes1 => primes1 ++ primes2 } }
f3 foreach { primes => println(primes.mkString(", ")) }
```

can be rewritten to

Advantage: Program structure is easier to understand.

Promises (1)

- Futures and promises represent two aspects of a single-assignment variable:
 - The promise allows you to assign a value to the future.
 - The future allows you to read the value.



Example usage of promises:

```
val p = Promise[String]()
p.future onComplete {
  case Success(v) => println(v)
  case Failure(ex) => println(ex)
}
p success "Hello"
// p.failure new Exception("failed")
```

Promises (2)

- A new value can only be assigned once to a future.
 - success, failed, and complete throw an exception if one tries to override a value.
 - trySuccess, tryFailed, and tryComplete return a value to indicate that the assignment was successful.
- It's easy to implement futures by means of promises:

```
import ExecutionContext.Implicits.global

def doInFuture[T](computation: => T): Future[T] =
   val p = Promise[T]()
   global.execute(() =>
        try
        val result = computation
        p success result
   catch
        case NonFatal(e) => p failure e

p.future
```

Future-Callback Bridge

- The control flow of callback-based APIs is not apparent from the code.
- The API dictates the control flow \rightarrow inversion of control.
- Promises can be used to bridge the gap between callback-based APIs and futures.

```
case class Event(value: Any)
@FunctionalInterface
trait EventListener:
  def onEvent(evt: Event)
```

```
class EventSource:
  def addEventListener(l: EventListener) = ...
```

```
def eventOccurred(): Future[Event] =
  val p = Promise[Event]()
  val source = new EventSource
  source.addEventListener((evt: Event) => p trySuccess evt)
  p.future
```

```
eventOccurred() foreach { evt => println(s"event [$evt] occurred.") }
```

Futures and Blocking

 If all threads of the thread pool are blocked, no further future computation can be started.

- Computation of future f will only start after 1000 ms.
- This situation is called thread starvation: shared resources are made unavailable for long periods by greedy threads.
- Avoid blocking whenever possible.
- If blocking is absolutely necessary, enclose code in a blocking call.

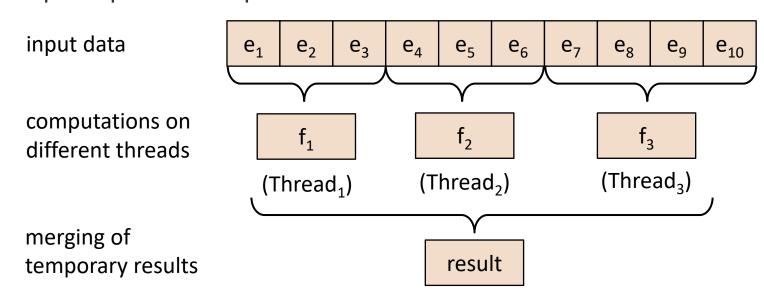
The execution context will provide additional worker threads if necessary.

Data Parallel Collections

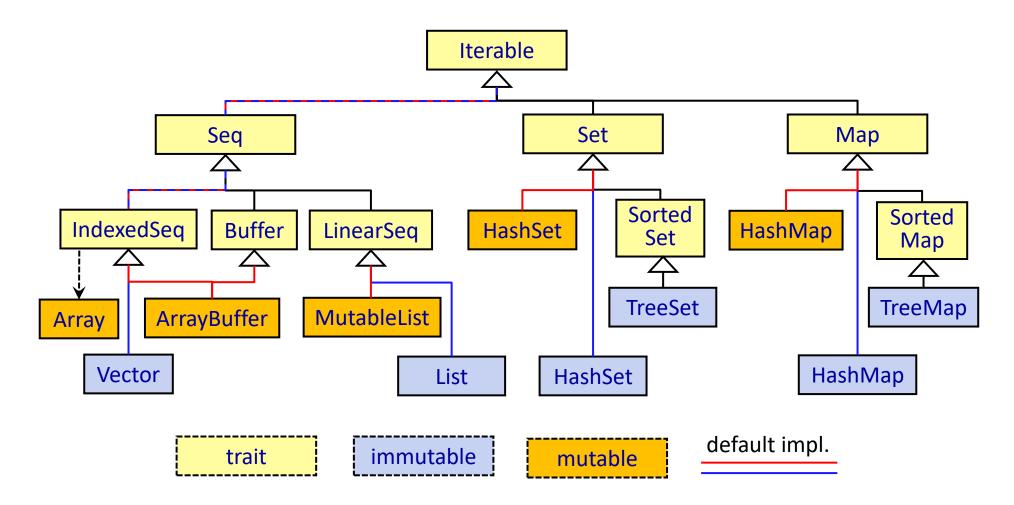
[Prokopec, 2017; p. 137 ff.]

Data Parallelism

- There are two forms of parallelism
 - Task parallelism (control parallelism) focusses on distributed tasks which are performed across different threads and processors.
 - Data parallelism focusses on performing the same operations on different data elements.
- Basic principle of data parallelism



Scala's Collection Hierarchy



Semantics of Scala Collections

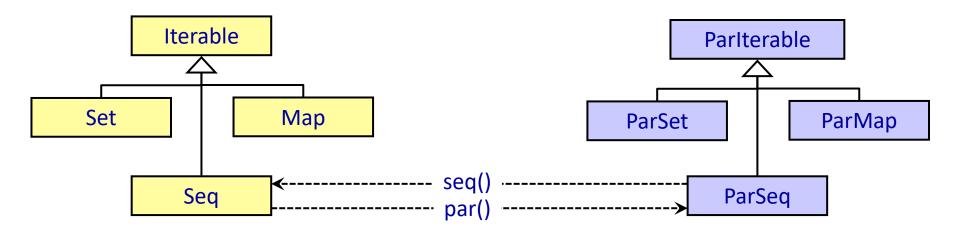
Iterable:

- Trait that implements behavior common to all collections based on the high-order function foreach.
- Provides an abstract method that generates an Iterator.

```
def foreach[U](f: Elem => U): Unit
def iterator: Iterator[T]
```

- Iterator defines abstract methods hasNext and next.
- Seq: Provides a method apply for indexing (numerical index).
- Buffer: Used to create sequences by appending (+=), prepending (+=:), inserting, updating, or removing elements.
- LinearSeq: Efficient implementation of head, tail, isEmpty. Most other operations are linear.
- Many types provide an apply method in the companion object that should be used to create collection instances (default implementation will be used).

Parallel Collections



- The method par converts a sequential collection into its parallel counterpart.
- Methods of parallel collections like map, foreach, filter, max, sum, etc. execute in parallel.
- Example:

```
def seqSumSquares(seq: Seq[Int]) = seq.map(i => i*i).sum

def parSumSquares(seq: ParSeq[Int]) = seq.map(i => i*i).sum

val numbers = 1 to 1000000

val seqSum = seqSumSquares(numbers)
val parSum = parSumSquares(numbers.par)
```

Specifics of Parallel Collection

- Parallelizable collections
 - Parallel splitters recursively split a collection into smaller parts:

```
trait Splitter[+T] extends Iterator[T]:
   def split: Seq[Splitter[T]]
```

- The asymptotic running time of split must be O(log n). Otherwise, the recursive division would produce too much overhead.
- Parallelizable collections: Array, Vector, HashSet, HashMap, etc.
- Non-parallelizable collections: List, Stream
- Side effects in parallel operations

```
var sum = new AtomicLong(0)
numbers.par foreach { i => sum += i }
```

- When multiple threads use variables, access must be synchronized, or atomic variables must be used.
- Parallel collection operation operations must be associative.