State

Chapter 6, with its progression of exercises, shows you how we can arrive at an elegant abstraction in library design. We are starting with imperative generator of random numbers (no exercises about this), then we make it pure (the RNG state). Then again we observe that using it is tedious so, we start to see the random number generators as state transformers (Rand) and then we remark that the main functions manipulating them can just as well manipulate any state transformers, not only random number generators (we arrive at the State abstraction).

All exercises are to be solved by extending the file State.scala. It contains several modules inside, to avoid name clashes between different abstractions. Once done, please hand in the completed State.scala file.

Exercise 1. Write a function that uses RNG.nextInt to generate a random integer between 0 and Int.maxValue. Make sure to handle the corner case when nextInt returns Int.MinValue, which does not have a non-negative counterpart.¹

```
def nonNegativeInt (rng: RNG): (Int, RNG)
```

Exercise 2. Write a function to generate a Double between 0 and 1, not including 1. Note: You can use Int.MaxValue to obtain the maximum positive integer value, and you can use x.toDouble to convert an x: Int to a Double.²

```
def double (rng: RNG): (Double, RNG)
```

Exercise 3. Write functions to generate (Int, Double) pairs and (Double, Int) pairs, where integers are non-negative. You should be able to reuse the functions you've already written. Add explicit return type annotations to both functions in your solution.³

```
1 def intDouble (rng: RNG)
2 def doubleInt (rng: RNG)
```

Exercise 4. Write a function to generate a list of random integers.⁴

```
def ints (count: Int) (rng: RNG)
```

Add explicit return type annotation to this function. Notice that all functions written so far have the same format: f[A]: RNG => (A, RNG), except that in the last case this type has been curried with one additional parameter. This motivates the generalization of the interface from now on to:

```
type Rand[A] =RNG =>(A, RNG)
```

See Section 6.4 in the text book for more details about the Rand[A] type. We shall develop an API for this type, like we did for Option. The API will allow us computing with random values, without explicitly carrying the generator state around.

Find the Rand type in the file and understand the unit and map implementations included, before you proceed.

¹Exercise 6.1 [Chiusano, Bjarnason 2014]

²Exercise 6.2 [Chiusano, Bjarnason 2014]

³Exercise 6.3 [Chiusano, Bjarnason 2014]

⁴Exercise 6.4 [Chiusano, Bjarnason 2014]

Exercise 5. Use map to reimplement double for Rand. See Exercise 2 above.

Observe how for Option we used the higher order API to avoid using pattern matching, and how here we use it to avoid being explicit about the state (and also to avoid decomposing, a.k.a. pattern matching, the results of random generators into value and new state).⁵

Exercise 6. Write the implementation of map2 based on the following signature. This function takes two generators, ra and rb, and calls a function f to combine the values produced by them:⁶

```
def map2[A,B,C] (ra: Rand[A], rb: Rand[B]) (f: (A, B) =>C): Rand[C]
```

Exercise 7. Implement sequence for combining a List of transitions into a single transition. Recall that we have already implemented sequence for Option—there is some experience to reuse here.

```
def sequence[A] (fs: List[Rand[A]]): Rand[List[A]]
```

Use sequence to reimplement the ints function you wrote before. For the latter, you can use the standard library function List.fill(n)(x) to make a list with x repeated n times.⁷

Exercise 8. Implement flatMap, and then use it to implement nonNegativeLessThan.8

```
def flatMap[A,B] (f: Rand[A]) (g: A =>Rand[B]): Rand[B]
```

Note: for Option we used map to compose a partial computation with a total computation. In here we used map to compose a random generator with a deterministic function. Similarly for flatMap. Function, Option. flatMap was used to compose two partial computations. On Rand the flatMap function is used to compose two random generators.

Exercise 9. Now we shall observe that everything we have done so far can just as well be done for other states, than RNG. Read beginning of Chapter 6.5, before solving this exercise.

Generalize the functions unit, map, map2, flatMap, and sequence. Note that this exercise is split into two parts of the .scala file (search for "Exercise 9" twice). Why is it split?

Exercise 10. We now connect the State and Streams. Recall from basics of computer science that automata and traces are intimately related: each automaton generates a language of traces. In our implementation automata are implemented using State and Streams can be used to represent traces.

Implement a function state2stream that given a state object and an initial state, produces a stream of values generated by this State object (this automaton).

This composition provides an alternative to using sequence suggested above.

Exercise 11. Use state2stream to generate a lazy stream of integer numbers. Finally, obtain a finite list of 10 random values from this stream.

Notice, how concise is the expression to obtain 10 random values from a generator using streams. This is because all our abstractions compose very well.

⁵Exercise 6.5 [Chiusano, Bjarnason 2014]

⁶Exercise 6.6 [Chiusano, Bjarnason 2014]

⁷Exercise 6.7 [Chiusano, Bjarnason 2014]

⁸Exercise 6.8 [Chiusano, Bjarnason 2014]