

# contextual: Simulating Contextual Multi-Armed Bandit Problems in R

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## Abstract

A large number of statistical decision problems in the social sciences and beyond can be framed as a (contextual) multi-armed bandit problem.

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A vignette for the [van Emden, Kaptein, and Postma \(2018\)](#) paper.

## 1. Introduction

In the canonical multi-armed bandit (MAB) problem a gambler faces a number of slot machines, each with a potentially different payoff. It is the gamblers goal to make as much profit (or, in the case of gambling, as little loss) as possible by sequentially choosing which machine to play, learning from the observations as she goes along.

## 2. Contextual Multi-Armed Bandits

In the canonical multi-armed bandit (MAB) problem a gambler faces a number of slot machines, each with a potentially different payoff. It is the gamblers goal to make as much profit (or, in the case of gambling, as little loss) as possible by sequentially choosing which machine to play, learning from the observations as she goes along.

## 3. Implementation of the contextual R package

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## 4. A basic example

```
bandit <- BasicBandit$new()  
bandit$set_weights(matrix(c(0.1, 0.9, 0.1, 0.5, 0.1, 0.1), 3, 3))  
policy <- EpsilonGreedyPolicy$new()
```

```
agent <- Agent$new(policy, bandit)
simulation <- Simulator$new(agent, horizon = 30L, simulations = 30L, worker_max = 1 )
context <- bandit$get_context()
history <- simulation$run()
```

## 5. Object orientation: extending contextual

The R6 package allows the creation of classes with reference semantics, similar to R's built-in reference classes. Compared to reference classes, R6 classes are simpler and lighter-weight, and they are not built on S4 classes so they do not require the methods package. These classes allow public and private members, and they support inheritance, even when the classes are defined in different packages.

One R6 class can inherit from another. In other words, you can have super- and sub-classes. Subclasses can have additional methods, and they can also have methods that override the superclass methods. In this example of a custom **contextual** bandit, we'll extend BasicBandit and override the initialize() method..

## 6. Special features

For instance, quantifying variance..

## 7. The art of optimal parallelisation

There is a very interesting trade of between the amount of parallelisation (how many cores, nodes used) the resources needed to compute a certain model, and the amount of data going to and fro the cores.

PERFORMANCE DATA —————

on 58 cores: k3\*d3 \* 5 policies \* 300 \* 10000 -> 132 seconds on 120 cores: k3\*d3 \* 5 policies \* 300 \* 10000 -> 390 seconds

on 58 cores: k3\*d3 \* 5 policies \* 3000 \* 10000 -> 930 seconds on 120 cores: k3\*d3 \* 5 policies \* 3000 \* 10000 -> 691 seconds

## 8. Extra greedy UCB

In the canonical multi-armed bandit (MAB) problem a gambler faces a number of slot machines, each with a potentially different payoff. It is the gamblers goal to make as much profit (or, in the case of gambling, as little loss) as possible by sequentially choosing which machine to play, learning from the observations as she goes along.

## 9. Conclusions

The goal of a data analysis is not only to answer a research question based on data but also

to collect findings that support that answer. These findings usually take the form of a table, plot or regression/classification model and are usually presented in articles or reports.

## 10. Acknowledgments

Thanks go to CCC.

## References

van Emden R, Kaptein M, Postma E (2018). *contextual: Simulating Contextual Multi-Armed Bandit Problems in R*. Jheronimus Academy of Data Science.

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