## Collective risk

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## Statement

The clients of an insurance company generate claims according to independent (homogeneous) Poisson processes with common rate  $\lambda$ . The claims are independent and the amount of each of them follows a distribution with cdf  $F(x) = 1 - (100/x)^{2.5}$  if x > 100. New clients enroll the company according to another (homogeneous) Poisson process with rate  $\nu$  and the time they stay in the company follows an exponential distribution with rate  $\mu$ . Each policyholder pays a fixed amount a every year to the company. If the policyholder stays in the company only during a fraction of the year, she will pay the corresponding fraction of a which is assumed to be paid on a continuous manner.

If the initial capital is  $c_0$  and the initial number of clients is  $n_0$ , we want to compute the probability that the capital of the company remains positive during some given time  $t_l$ .

## Introduction

Each policyholder has a fixed amount a to pay every year, so the capital of the company will increase by a times the number of clients every fraction of year.

On the other hand, they also hold the right to claim the premium on the policy, which happens according to a Poisson process with common rate  $\lambda$ . The amount of is this premium is distributed according to a Pareto distribution with parameters  $\alpha = 2.5$  and  $\beta = 100$ .

The clients of the company can leave the company at any time, and the time they stay in the company follows an exponential distribution with rate  $\mu$ .

Clients come according to an independent (homogeneous) Poisson processes with rate  $\nu$ , and they stay according to an exponentially distributed time with rate  $\mu$ .

So in general, the capital of the company at any time t will be:

$$C(t) = c_0 + at(n_0 + N_A(t) - N_D(t)) - \sum_{j=1}^{N_C(t)} X_j$$

where  $N_A(t)$  is the number of clients that arrive by time t,  $N_D(t)$  is the number of clients that leave by time t,  $N_C(t)$  is the number of claims that arrive by time t,  $X_j$  is the amount of the j-th claim, and n(t) is the number of clients at time t.

## The project

Project development including the code and emphasizing the main difficulties and most important parts.

First, to simulate the amount of the claims we will define a function with the inverse of the cdf of the Pareto distribution. Mathematically, this function is:

$$F^{-1}(x) = \frac{\beta}{\sqrt[\alpha]{(1-x)}} = \frac{100}{\sqrt[2.5]{(1-x)}}$$

```
inverse.claim <- function(x) {
  return(max((100 / (1 - x)^(1 / 2.5)), 0))
}</pre>
```

The algorithm we will follow is, for each path:

- 1. Initialize variables t = 0, n = n0, c = c0.
- 2. Compute the time of the first event: dt = rexp(1, lambda \* n + mu \* n + nu)
- 3. Update the time: t = t + dt
- 4. While t < tl:
  - 4.1. Update capital with the proportional part of the payments: c = c + n \* a \* dt
  - 4.2. Decide the type of event and update in each case:
    - 4.2.1. If  $u < \lambda * n/(\lambda * n + \mu * n + \nu)$ , claim event so update capital  $= c X_j$
    - 4.2.2. If  $u < (\lambda * n + \mu * n)/(\lambda * n + \mu * n + \nu)$ , departure event so n = n 1
    - 4.2.3. Else, enrollment event, so n = n + 1
  - 4.3. Compute the time of the next event: dt = rexp(1, lambda \* n + mu \* n + nu)
  - 4.4. Update the time: t = t + dt

Thus for one simulation we have:

```
simulate_one_insurance <- function(c0, n0, a, t1, lambda, mu, nu) {
    # Initialize variables
    t <- 0
    n <- n0
    c <- c0
    # Update capital with annuity payments
    dt <- rexp(1, lambda * n + mu * n + nu)
    t <- t + dt
    while (t < t1) {
        # Update the capital with the proportional part of the payments
        c <- c + n * a * dt
        # Decide the type of event</pre>
```

```
u <- runif(1)
    if (u < lambda * n / (lambda * n + mu * n + nu)) {</pre>
      # Claim event
      claim_amount <- inverse.claim(runif(1))</pre>
      c <- c - claim_amount</pre>
    } else if (u < (lambda * n + mu * n) / (lambda * n + mu * n + nu)) {
      # Departure event
     n \leftarrow max(n - 1, 0)
    } else {
      # Enrollment event
      n < -n + 1
    # Stop if capital becomes negative
    if (c < 0) break
    # Time of next event
    dt \leftarrow rexp(1, lambda * n + mu * n + nu)
    t <- t + dt
 }
 return(c)
}
And for multiple:
simulate_insurance <- function(c0, n0, a, t1, lambda, mu, nu, MC) {</pre>
  final_capitals <- vector("numeric", MC)</pre>
  # Simulate MC paths
  final_capitals <- replicate(MC, simulate_one_insurance(c0, n0, a, t1, lambda, mu, nu))</pre>
  # Compute and return outputs
  fraction <- sum(final_capitals > 0) / MC
  pos fc <- final capitals[final capitals > 0]
  mean_final_capitals <- mean(pos_fc)</pre>
  sd_final_capitals <- sd(pos_fc)</pre>
  p1 <- ggplot() +
        geom_line(mapping= aes(x = 1:length(pos_fc),
                                y = cumsum(pos_fc)/(1:length(pos_fc)))) +
        labs(x = "Number of simulations", y = "Mean of final capitals",
               title = "Convergence of the mean of the final capitals")
  plot(p1)
  p2 <- ggplot() +
        geom_histogram(mapping= aes(x = pos_fc, y=after_stat(density)),
                        binwidth=5000, color = "black", fill = "white") +
        stat_function(fun=dnorm, args=list(mean=mean_final_capitals, sd = sd_final_capitals),
                       color = "red") +
        labs(x = "Final capitals", y = "Density",
               title = "Density of the final capitals")
  plot(p2)
```

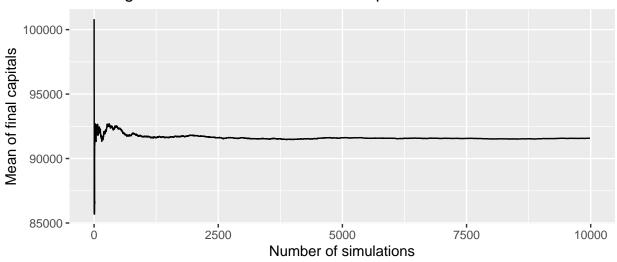
## Results

}

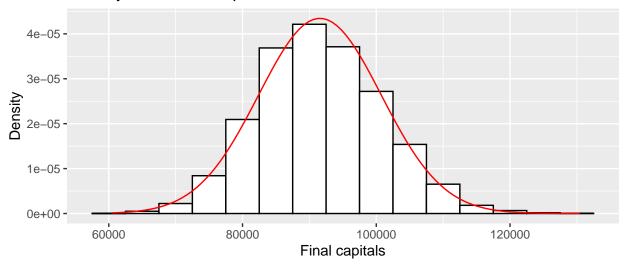
return(list(fraction = fraction, mean\_final\_capital = mean\_final\_capitals,

sd\_final\_capital = sd\_final\_capitals))

# Convergence of the mean of the final capitals



# Density of the final capitals



```
res
```

```
## $fraction
## [1] 0.9984
##
## $mean_final_capital
## [1] 91573.45
##
## $sd_final_capital
## [1] 9182.815
```

Numerical results, quality assessment of the approximations, and time efficiency of the algorithms.

# Conclusions

About the results, how the difficulties were solved, and possible alternative approaches. Keep the focus, the conclusions must be as brief as possible.

## References

Including textbooks, webpages, and class notes.