

Projeto Computational 1

Métodos Computacionais em Finanças

2024/2025

Afonso da Conceição Ribeiro (ist1102763)

Mestrado em Engenharia e Ciência de Dados

Instituto Superior Técnico – Universidade de Lisboa

Índice

1.		2
(a)	2
(b)	5
2.		7

1.

(a)

Let $V(S, t)$ be the price of an option with underlying price S at calendar time $t \in [0, T]$. For all vanilla contracts (European/American, call/put), V satisfies the Black-Scholes partial differential equation

$$\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} S^2 \frac{\partial^2 V}{\partial S^2} + r S \frac{\partial V}{\partial S} - r V = 0, \quad 0 < S < S^*, \quad 0 \leq t \leq T, \quad (1)$$

subject to terminal and boundary data that depend on the contract type. To turn the terminal-value problem (1) into an initial-value problem we introduce the forward time

$$t = T - t, \quad U(S, t) := V(S, T - t).$$

(also with t to simplify the notation).

Then U obeys

$$\begin{cases} \frac{\partial U}{\partial t} = \frac{\sigma^2}{2} S^2 \frac{\partial^2 U}{\partial S^2} + r S \frac{\partial U}{\partial S} - r U, & 0 < S < S^*, \quad 0 < t < T \\ U(S, 0) = u_0(S), & 0 \leq S \leq S^*, \\ U(0, t) = u_a(t), & 0 \leq t \leq T, \\ U(S^*, t) = u_b(t), & 0 \leq t \leq T. \end{cases} \quad (2)$$

For each contract, the functions u_0, u_a, u_b are assumed known.

We discretise the spatial axis $S \in [0, S^*]$ with a uniform grid

$$S_i = i h_S, \quad h_S = \frac{S^*}{N_S}, \quad i = 0, \dots, N_S,$$

and approximate the equation's derivatives in space at a certain point S_i for a fixed time, yielding, for the interior nodes $i = 1, \dots, N_S - 1$,

$$\frac{\partial U}{\partial t}(S_i, t) = U(S_{i-1}, t) \underbrace{\left(\frac{\sigma^2}{2} i^2 - \frac{r i}{2}\right)}_{:= \alpha_i} + U(S_i, t) \underbrace{(-\sigma^2 i^2 - r)}_{:= \beta_i} + U(S_{i+1}, t) \underbrace{\left(\frac{\sigma^2}{2} i^2 + \frac{r i}{2}\right)}_{:= \gamma_i}.$$

Collecting the interior unknowns into the vector

$$W(t) := [U_1(t), \dots, U_{N_S-1}(t)]^\top,$$

we obtain the ODE system

$$W'(t) = A W(t) + b(t), \quad (3)$$

where $A \in \mathbb{R}^{(N_S-1) \times (N_S-1)}$ is tridiagonal with band entries $\alpha_i, \beta_i, \gamma_i$ and $b(t)$ encodes the (possibly time-dependent) boundary values $u_a(t)$ and $u_b(t)$.

Finally, we use the fourth order Runge-Kutta method:

$$\begin{cases} f_1 = h F(t, W), \\ f_2 = h F(t + \frac{h}{2}, W + \frac{f_1}{2}), \\ f_3 = h F(t + \frac{h}{2}, W + \frac{f_2}{2}), \\ f_4 = h F(t + h, W + f_3), \\ W(t + h) = W(t) + \frac{1}{6}(f_1 + 2f_2 + 2f_3 + f_4), \end{cases}$$

with $F(t, W) = A W + b(t)$.

Input and output of MOL_RK4

Input: r risk-free rate [year^{-1}];
 σ volatility [$\text{year}^{-1/2}$];
 T maturity (calendar time horizon);
 S^* truncation of the S -domain;
 N_S, N_t spatial and temporal resolutions;
 $u_0(S)$ payoff function $U(S, 0)$;
 $u_a(t)$ left boundary $U(0, t)$;
 $u_b(t)$ right boundary $U(S^*, t)$.

Output: A Matlab *struct* `out = { .S, .t, .U }`, where

- `.S` the $(N_S + 1)$ -vector of spatial nodes S_i ;
- `.t` the $(N_t + 1)$ -vector of forward times t ;
- `.U` the $(N_S + 1) \times (N_t + 1)$ array with entries $U_{i,j} \approx U(S_i, t_j)$.

```

1 function out = MOL_RK4(r, sigma, T, S_star, NS, Nt, u_0, u_a, u_b)
2     %% Spatial grid
3     hS = S_star / NS;
4     h = T / Nt;
5     S = 0 : hS : S_star;
6     i = (1:NS-1)';
7
8
9     %% Finite-difference coefficients  $\alpha_i, \beta_i, \gamma_i$ 
10    alpha_i = 0.5*sigma^2 .* i.^2 - 0.5*r .* i;
11    beta_i = -sigma^2 .* i.^2 - r;
12    gamma_i = 0.5*sigma^2 .* i.^2 + 0.5*r .* i;
13
14    A = spdiags([alpha_i beta_i gamma_i], [-1 0 1], NS-1, NS-1); % A_ML
15
16
17    %% Initial condition W(0) (pay-off)
18    W = u_0(S(2:NS)).';
19
20
21    %% Pre-allocate storage for all slices
22    U = zeros(NS+1, Nt+1);
23    U(:,1) = u_0(S).'; % (t=0 forward)
24    b = zeros(NS-1,1); % b_ML(t)
25
26
27    %% RK-4 loop (forward time)
28    for n = 0:Nt-1
29        t = n * h;
30
31        % Boundary values V(S=0,t) and V(S=S*,t)
32        V_left = u_a(t); % U(0,t)
33        V_right = u_b(t); % U(S*,t)
34        b(1) = alpha_i(1) * V_left;
35
36        % Runge-Kutta stages
37        f1 = h * (A*W + b);
38

```

```
39     t_half    = t + 0.5*h;
40     V_left_h  = u_a(t_half);
41     b(1)      = alpha_i(1) * V_left_h;
42     f2 = h * (A*(W + 0.5*f1) + b);
43     f3 = h * (A*(W + 0.5*f2) + b);
44
45     t_next    = t + h;
46     V_left_n  = u_a(t_next);
47     b(1)      = alpha_i(1) * V_left_n;
48     f4 = h * (A*(W + f3) + b);
49
50     % RK-4 update
51     W = W + (f1 + 2*f2 + 2*f3 + f4)/6;
52
53     % Assemble full vector V(S_i, t_next) and store
54     U(:,n+2) = [V_left_n ; W ; u_b(t_next)];
55 end
56
57
58 %% Output
59 out.S = S;
60 out.t = 0:h:T;
61 out.U = U;
62 end
```

(b)

Specification for a European put

For a European put with strike K , the data functions are

$$u_0(S) = \max(K - S, 0), \quad u_a(t) = K e^{-rt}, \quad u_b(t) = 0.$$

The routine sets the parameters r, σ, T, K, S^* and the grids (N_S, N_t) , constructs function handles for u_0, u_a, u_b , calls `MOL_RK4`, and finally produces

- a 2D slice $V(S, 0) = U(S, T)$ at calendar time $t = 0$, and
- a 3D surface $V(S, t) = U(S, T - t)$ for $0 \leq t_{\text{cal}} \leq T$.

```

1 %% Parameters
2 r      = 0.06;
3 sigma  = 0.3;
4 T      = 1;
5 K      = 10;
6 S_star = 15;
7
8 NS = 400;
9 Nt = 13000;
10
11 %% Functions for European put option
12 u_0 = @(S) max(K-S,0);
13 u_a = @(t) K*exp(-r*t);
14 u_b = @(t) 0*t;
15
16 %% Run
17 sol = MOL_RK4(r,sigma,T,S_star,NS,Nt,u_0,u_a,u_b);
18
19 %% 2D price at calendar time t=0 (last column)
20 V_today = sol.U(:,end);
21 figure;
22 plot(sol.S, V_today), grid on
23 title('European put V(S,0)')
24
25 %% 3D surface V(S,t)
26 [Tgrid, Sgrid] = meshgrid(T - sol.t, sol.S); % calendar time axis
27 figure;
28 mesh(Tgrid, Sgrid, sol.U)
29 xlabel('t'),
30 ylabel('S'),
31 zlabel('V(S,t)')
32 title('European put option value, V(S,t)')
33 view(135,30);

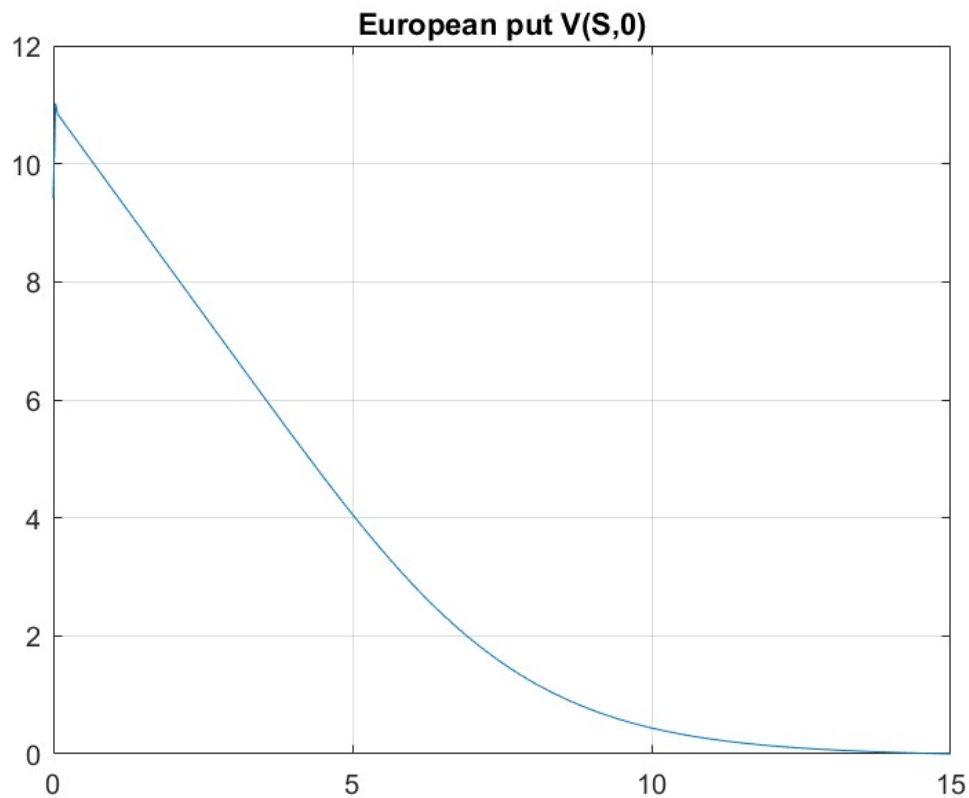
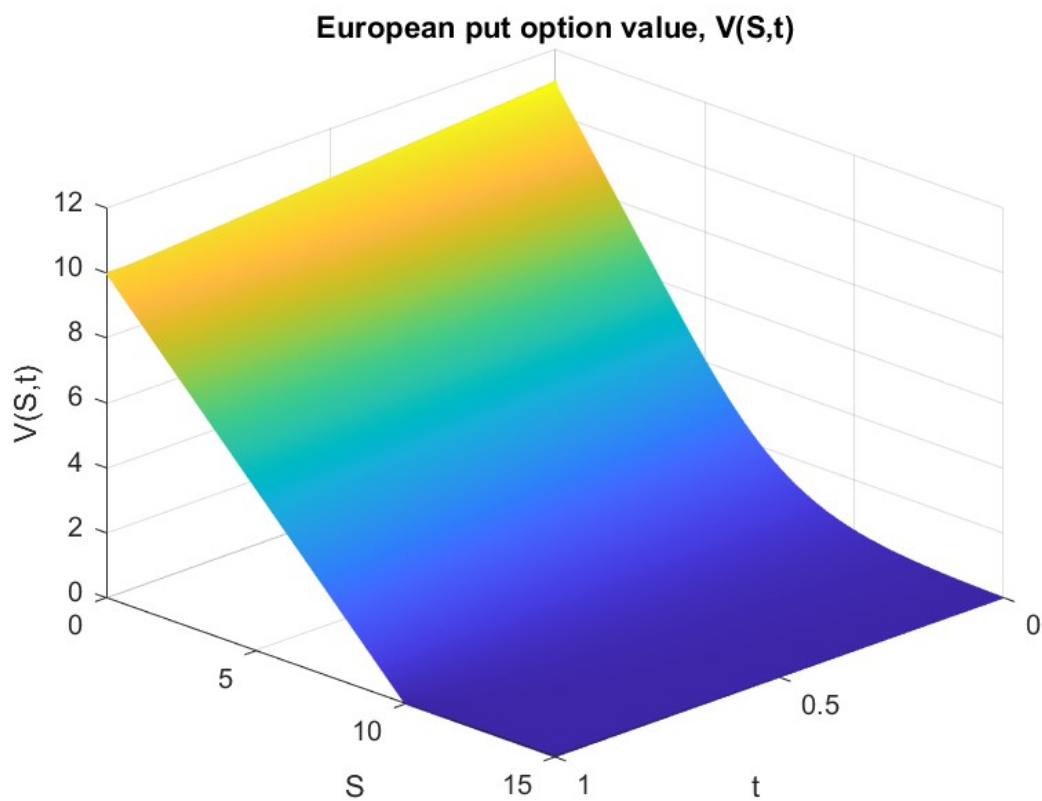
```

Discussion of the numerical results

With the default grid $(N_S, N_t) = (400, 13000)$ the computation achieves $|V_{\text{num}} - V_{\text{BS}}| < 10^{-3}$ for all $S \in [0, 15]$, when compared with the closed-form Black-Scholes price.

Figure 1 shows the option price at today's date ($t = 0$). The curve starts at $V(0, 0)$, is decreasing and convex, and approaches 0 for $S \gg K$, exactly as theory predicts.

Figure 2 depicts the full surface $V(S, t)$. At $t = T$, the surface equals the kinked pay-off $\max(K - S, 0)$; as t decreases, the surface becomes smooth in S .

Figura 1: European put value $V(S,0)$ Figura 2: Surface $V(S,t)$ for $0 \leq t \leq T$ (calendar time axis)

2.

Let $V(S, t)$ be the price of an American put option with underlying price S at time $t \in [0, T]$. It satisfies the Black-Scholes inequality, which takes the form of a variational inequality:

$$\min \left\{ V(S, t) - (K - S)^+, \frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV \right\} = 0, \quad (4)$$

subject to the boundary conditions

$$V(0, t) = K, \quad V(S^*, t) = 0, \quad V(S, T) = \max(K - S, 0),$$

for all $t \in [0, T]$ and some large $S^* > K$.

To solve (4) numerically, we discretise the time and asset axes:

$$S_i = i h_S, \quad h_S = \frac{S^*}{N_S}, \quad t_n = n h_t, \quad h_t = \frac{T}{N_t},$$

for $i = 0, \dots, N_S$, $n = 0, \dots, N_t$. Let V_i^n denote the approximation to $V(S_i, t_n)$.

We apply the Crank-Nicolson scheme to the PDE part in (4), with central differences in space:

$$\begin{aligned} \alpha_i &= \frac{1}{4} h_t (\sigma^2 i^2 - r i), \\ \beta_i &= -\frac{1}{2} h_t (\sigma^2 i^2 + r), \\ \gamma_i &= \frac{1}{4} h_t (\sigma^2 i^2 + r i). \end{aligned}$$

These define a tridiagonal matrix A acting on the interior points $V_1^n, \dots, V_{N_S-1}^n$. The Crank-Nicolson system at each time step reads

$$BV^n = d^{n+1}, \quad B = I - \frac{1}{2} h_t A, \quad d^{n+1} = \left(I + \frac{1}{2} h_t A \right) V^{n+1} + BC.$$

However, due to the early exercise constraint $V_i^n \geq \max(K - S_i, 0)$, this becomes a linear complementary problem. We solve it with the PSOR method:

$$V_i^{(k+1)} = \max \left\{ \max(K - S_i, 0), V_i^{(k)} + \omega \frac{r_i^{(k)}}{B_{ii}} \right\},$$

where $r_i^{(k)}$ is the residual at iteration k .

Input and output of CN_PSOR

Input: r risk-free rate;
 σ volatility;
 T maturity;
 K strike price;
 S^* truncation of the S -domain;
 N_S, N_t grid resolutions.

Output: A Matlab *struct* $\text{out} = \{ .S, .t, .V \}$, where

$.S$ the $(N_S + 1)$ -vector of asset prices S_i ;
 $.t$ the $(N_t + 1)$ -vector of times t_n ;
 $.V$ the $(N_S + 1) \times (N_t + 1)$ matrix of values V_i^n .

```

1 function [S_grid, t_grid, V] = CN_PSOR(r, sigma, T, K, S_star, NS, Nt)
2     %% Grids
3     dS      = S_star / NS;
4     dt      = T / Nt;
5     S_grid = linspace(0, S_star, NS+1)';
6     t_grid = linspace(0, T, Nt+1);
7
8     %% Grid indices
9     I = 2:NS;
10    S = S_grid(I);
11    M = NS - 1;
12
13    %% Coefficients for tridiagonal matrices
14    i = (1:M)'; % i indexes interior points
15    alpha = 0.25 * dt * (sigma^2 * i.^2 - r * i);
16    beta  = -0.5 * dt * (sigma^2 * i.^2 + r);
17    gamma = 0.25 * dt * (sigma^2 * i.^2 + r * i);
18
19    main_diag = 1 - beta;
20    lower_diag = -alpha(2:end);
21    upper_diag = -gamma(1:end-1);
22
23    %% Matrix B (I - dt/2*A)
24    B = spdiags([[lower_diag; 0], main_diag, [0; upper_diag]], -1:1, M, M);
25
26    %% Matrix A+ (I + dt/2*A)
27    main_plus = 1 + beta;
28    low_plus = alpha(2:end);
29    up_plus = gamma(1:end-1);
30    A_plus = spdiags([[low_plus; 0], main_plus, [0; up_plus]], -1:1, M, M);
31
32    %% Payoff and boundary conditions
33    V = zeros(NS+1, Nt+1);
34    V(:, end) = max(K - S_grid, 0);
35    V(1, :) = K; V(end, :) = 0;
36    payoff = max(K - S, 0);
37
38    %% Relaxation
39    omega = 1.3;
40    tol = 1e-7;
41    max_iter = 500;

```



```

42
43 %% Time-stepping loop
44 for n = Nt:-1:1
45     rhs = A_plus * V(I, n+1);
46
47     % Boundary conditions
48     rhs(1) = rhs(1) + alpha(1) * (V(1, n+1) + V(1, n));
49     rhs(end) = rhs(end) + gamma(end) * (V(end, n+1) + V(end, n));
50
51     % Initial guess = last solution
52     x = V(I, n+1);
53
54     % PSOR iterations
55     for it = 1:max_iter
56         x_old = x;
57         for i = 1:M
58             if i == 1
59                 residual = rhs(i) - B(i,i)*x(i) - B(i,i+1)*x(i+1);
60             elseif i == M
61                 residual = rhs(i) - B(i,i-1)*x(i-1) - B(i,i)*x(i);
62             else
63                 residual = rhs(i) - B(i,i-1)*x(i-1) - B(i,i)*x(i) - B(i,
64 i+1)*x(i+1);
65             end
66             x(i) = max(payoff(i), x(i) + omega * residual / B(i,i));
67         end
68         if norm(x - x_old, inf) < tol
69             break;
70         end
71     end
72     V(I, n) = x;
73 end
74 end

```

Numerical results for the American put

We set the parameters

$$r = 0.06, \quad \sigma = 0.3, \quad T = 1, \quad K = 10, \quad S^* = 15,$$

and use the grid $(N_S, N_t) = (400, 1000)$.

Figure 3 shows the value of the option at three calendar times $t = 0, 0.5, 1$. At $t = T$ the price matches the payoff $\max(K - S, 0)$, while for $t < T$ the value lies strictly above it in the continuation region.

Figure 4 shows the continuation region: the values $V(S, t)$ that are strictly above the payoff. This surface is smooth in both S and t and bounded below by the free boundary $S_f(t)$.

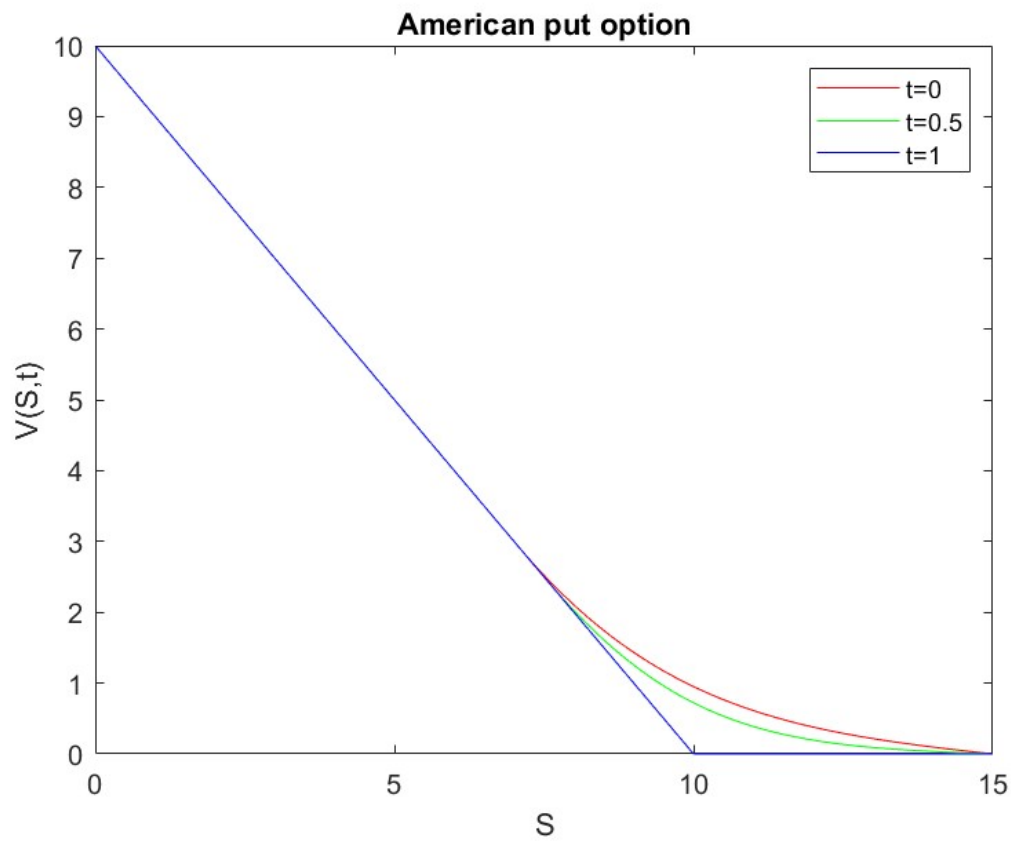


Figura 3: American put value $V(S, t)$ for $t = 0, 0.5, 1$

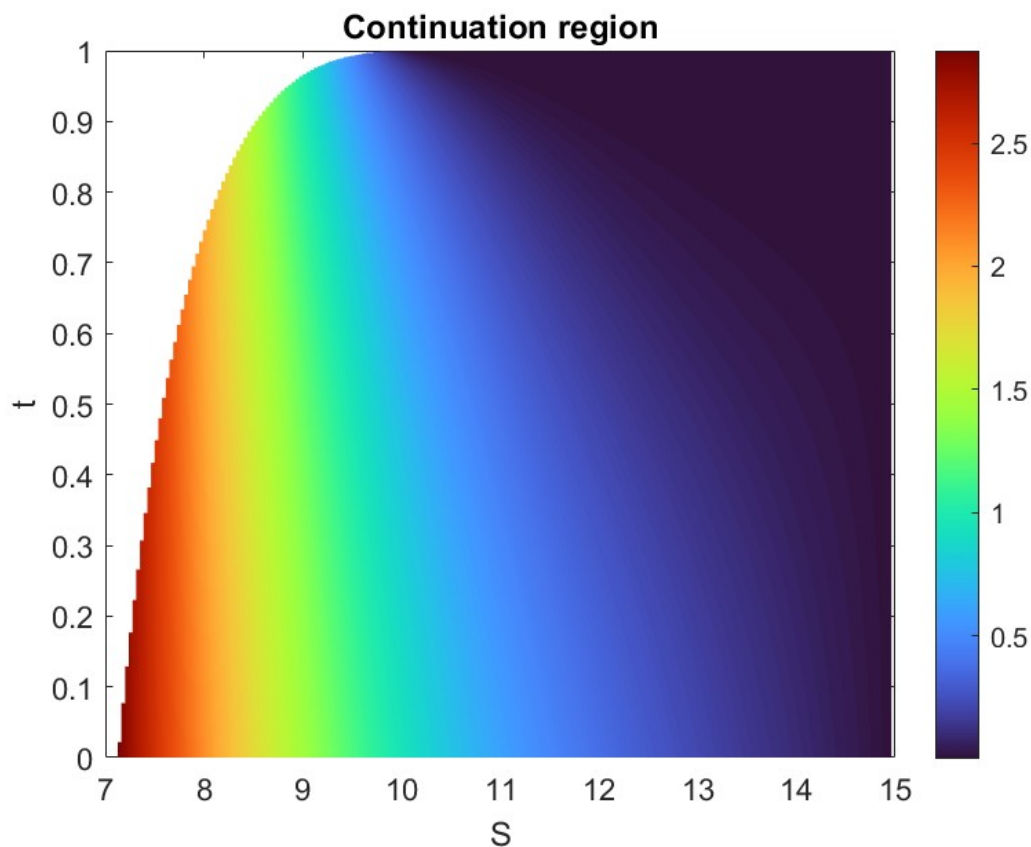


Figura 4: Continuation region $V(S, t) > (K - S)^+$