

# Greičiausiojo nusileidimo (*angl. steepest descent*) metodo C (C++) realizacija

# Greičiausiojo nusileidimo metodo realizacija

- Iš [http://www.mymathlib.com/optimization/nonlinear/unconstrained/steepest\\_descent.html](http://www.mymathlib.com/optimization/nonlinear/unconstrained/steepest_descent.html)
- Parsisiunčiame *steepest\_descent.c*

# *steepest\_descent.c* (I)

```
////////////////////////////////////  
// File: steepest_descent.c  
// Routines:  
// Steepest_Descent  
////////////////////////////////////  
#include <stdlib.h> // required for malloc(), free(), and NULL  
#include <stdio.h>  
// Externally Defined Routines  
  
double Vector_Max_Norm(double v[], int n);  
  
int Minimize_Down_the_Line(double (*f)(double *), double x[], double fx,  
                           double *p, double v[], double y[], double cutoff,  
                           double cutoff_scale_factor, double tolerance, int n);  
  
void Copy_Vector(double *d, double *s, int n);
```

## *steepest\_descent.c* (2)

```
////////////////////////////////////  
// int Steepest_Descent(double (*f)(double *), //  
// void (*df)(double *, double *), //  
// int (*stopping_rule)(double*, double, double*, double, double*, //  
// int, int), //  
// double a[], double *fa, double *dfa, double cutoff, //  
// double cutoff_scale_factor, double tolerance, int n) //  
// //  
// Description: //  
// Given a nonlinear function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ , the optimal steepest descent //  
// method searches for the minimum of  $f$  by starting at a predetermined //  
// point  $x$  and then following the line from  $x$  in the direction opposite //  
// to that given by  $\text{grad } f$  evaluated at  $x$  until a minimum on the line is //  
// reached. The process is repeated until halted by the stopping rule. //  
// The steepest descent method is best used when the point  $x$  is far away //  
// from the minimum. Near the minimum, the norm of  $\text{grad } f$  is small and //  
// convergence is slow and another method should be used accelerate //  
// convergence to the final approximation of the minimum. //  
// //  
// It is possible that if the initial point happens to be the location of //  
// a local maximum or a saddle point, the gradient will vanish and the //  
// procedure will fail. //  
// It is also possible that the method will converge to a saddle point //  
// so that the result should be checked and if it is a saddle point then //  
// try a different initial point.
```

## *steepest\_descent.c* (3)

```
// Arguments: //
// double (*f)(double *) //
//     Pointer to a user-defined function of a real n-dimensional vector //
//     returning a real number (type double). //
// void (*df)(double *, double *) //
//     Pointer to a user-defined function which returns the gradient of the //
//     function f, above, in the second argument evaluated at the point //
//     in the first argument. I.e. df(x,dfx), where dfx = grad f(x). //
// int (*stopping_rule)(double*, double, double*, double, double*, int, //
//     int) //
//     Pointer to a user-defined function which controls the iteration of //
//     the steepest descent method. If the stopping rule is non-zero the //
//     process is halted and if zero then the process continues iterating. //
//     Steepest descent converges slowly in the neighborhood of an //
//     extremum and another method is then called to converge to the final //
//     solution. //
//     The arguments are: original point, the value of the function f //
//     evaluated at the original point, the new point, the value of the //
//     function evaluated at the new point, the gradient of the function //
//     at the new point, the total number of iterations performed and the //
//     dimension, n, if a point and the gradient. //
```

# Mano (user-defined) funkcijos

```
// Apskaiciuoja Six-hump Camel Back funkcijos reiksme taske x
double SixHumpCamelBack(double *x){
    return (4-2.1*x[0]*x[0]+x[0]*x[0]*x[0]*x[0]/3)*x[0]*x[0] + x[0]*x[1] +
        (-4+4*x[1]*x[1])*x[1]*x[1];
}

// Apskaiciuoja Six-hump Camel Back gradiento reiksme taske x
void SixHumpCamelBackGradient(double *x, double *fGrad){
    fGrad[0] = 8*x[0]-8.4*x[0]*x[0]*x[0]+2*x[0]*x[0]*x[0]*x[0]*x[0]+x[1];
    fGrad[1] = x[0]-8*x[1]+16*x[1]*x[1]*x[1];
}

// Algoritmo sustojimo salygas kontroliuojanti funkcija
int StoppingRule(double* a, double fa, double* x, double fx, double* dfa, int
iteration, int n){
    double fEps = abs(fx - fa); // Funkcijos reiksmiu skirtumas
    double xa[n];
    for(int i = 0; i < n; ++i) xa[i] = x[i]-a[i];
    double xEps = Vector_Max_Norm(xa, 2); // Argumento skirtumo norma
    double dfaEps = Vector_Max_Norm(dfa, 2); // Gradiento norma
    iteration++; // Iteraciju skaitiklis
    if(iteration > 3)
        return -6;
    else
        return 0;
}
```

## *steepest\_descent.c* (4)

```
//      double *a                                     //
//      On input, a is the initial point to start the descent.  On output,` //
//      a is the final point before the iteration is halted.      //
//      double *fa                                       //
//      On input and output, *fa is the value of f(a[]).         //
//      double *dfa                                       //
//      Working storage used to hold the gradient of f at a. Should be //
//      dimensioned at least n in the calling routine.           //
//      double cutoff                                     //
//      The upper bound of the parameter p during the line search where //
//      the line is  $x - pv$ , v is the gradient of f at x,  $0 \leq p \leq \text{cutoff}$ . //
//      double cutoff_scale_factor                       //
//      A parameter which limits the displacement in any single step during //
//      the parabolic extrapolation phase of the line search.     //
//      double tolerance                                  //
//      A parameter which controls the termination of the line search. //
//      The line search is terminated when the relative length of the //
//      interval of uncertainty to the magnitude of its mid-point is less //
//      than or equal to tolerance.  If a nonpositive number is passed, //
//      tolerance is set to sqrt(DBL_EPSILON).               //
//      int      n                                       //
//      On input, n is the dimension of a and dfa.
```

## *steepest\_descent.c* (5)

```
// Return Values: //
// 0 Success //
// -1 In the line search three points are collinear. //
// -2 In the line search the extremum of the parabola through the three //
// points is a maximum. //
// -3 In the line search the initial points failed to satisfy the //
// condition that  $x_1 < x_2 < x_3$  and  $fx_1 > fx_2 < fx_3$ . //
// -4 Not enough memory. //
// -5 The gradient evaluated at the initial point vanishes. //
// Other user specified return from stopping_rule(). //
// //
// Example: //
// extern double f(double*); //
// extern void df(double*, double*); //
// extern int Stopping_Rule(double*, double, double*, double, double*, //
// int, int); //
// #define N //
// double cutoff, cutoff_scale_factor, tolerance; //
// double a[N], dfa[N]; //
// //
// (your code to initialize the vector a, set cutoff, ) //
// (cutoff_scale_factor, and tolerance. ) //
// //
// err = Steepest_Descent( f, df, Stopping_Rule, a, fa, dfa, cutoff, //
// cutoff_scale_factor, tolerance, N); //
// //
////////////////////////////////////
```



## *steepest\_descent.c* (6)

```
int Steepest_Descent(double (*f)(double *), void (*df)(double *, double *),
    int (*stopping_rule)(double*, double, double*, double, double*, int, int),
    double a[], double *fa, double *dfa, double cutoff,
    double cutoff_scale_factor, double tolerance, int n)
{
    double* x;
    double* initial_a = a;
    double* tmp;
    double fx;
    double p;
    int err = 0;
    int iteration = 0;

    x = (double *) malloc(n * sizeof(double));
    if ( x == NULL ) { err = -3; }

    df(a, dfa);
    if (Vector_Max_Norm(dfa,n) == 0.0) err = -5;
```

## *steepest\_descent.c* (7)

```
while (!err) {
    if (Vector_Max_Norm(dfa,n) == 0.0) break;
    err = Minimize_Down_the_Line(f, a, *fa, &p, dfa, x, cutoff,
                                cutoff_scale_factor, tolerance, n);

    if ( err ) break;
    fx = f(x);
    iteration++;
    df(x, dfa);
    err = stopping_rule( a, *fa, x, fx, dfa, iteration, n);
    *fa = fx;
    tmp = a;
    a = x;
    x = tmp;
}
if (a != initial_a ) {
    Copy_Vector(initial_a, a, n);
    x = a;
}
free(x);
return err;
}
```

# *main.cpp* (I)

```
int main(int argc, const char * argv[])
{
    double region[] = {-1.9, 1.9, -1.1, 1.1};
    double a[N] = {0.0, 1.0}; // N-matis Vektorius
    /*srand(time(0)); // Naudoja vis kita seed'a
    double a[N]; // N-matis Vektorius
    for(int i = 0; i < N; ++i){
        a[i] = GetRandomNumber(region[2*i], region[2*i+1]);
    }*/
    double fa = SixHumpCamelBack(a); // Funkcijos reiksme pradiniam taske a
    double dfa[N];
    SixHumpCamelBackGradient(a, dfa); // Funkcijos gradiento reiksme taske a
    double cutoff = 1.0, cutoff_scale_factor = 1.0; // Pap. parametrai
    double tolerance = 0.01;
    int err = Steepest_Descent( SixHumpCamelBack, SixHumpCamelBackGradient, StoppingRule,
    a, &fa, dfa, cutoff, cutoff_scale_factor, tolerance, N);
```

## *main.cpp* (2)

```
switch (err)
{
    case 0:
        cout << "Success" << endl;
        break;
    case -1:
        cout << "In the line search three points are collinear." << endl;
        break;
    case -2:
        cout << "In the line search the extremum of the parabola through the
three points is a maximum." << endl;
        break;
    case -3:
        cout << "Int the line search the initial points failed to satisfy
the condition that  $x_1 < x_2 < x_3$  and  $fx_1 > fx_2 < fx_3$ ." << endl;
        break;
    case -4:
        cout << "Not enough memory." << endl;
        break;
    case -5:
        cout << "The gradient evaluated at the initial point vanishes." << endl;
    case -6:
        cout << "Exceed maximal number of iterations." << endl;
        break;
}
cout << "Greiciausio nusileidimo (angl. Steepest Descent) metodu" << endl;
cout << "surastas sprendinys yra:" << endl;
cout << "xMin = (" << a[0] << ", " << a[1] << ")" << endl;
cout << "f(xMin) = " << fa << endl;
return 0;
}
```

# Kartotinių paieškų metodas (*angl. multi start*)

1. Papildyti Jūsų turimas atsitiktinės paieškos (*angl. pure random search or Monte Carlo*) programas, kad jos iš 3 geriausių surastų sprendinių atliktų lokalias paieškas - sprendinių patikslinimus naudojantis greičiausio nusileidimo metodu (*angl. steepest descent*).

`git clone https://github.com/Remziukas/SteepestDescent.git`

2. Papildyti dar (bent) vienu laisvai pasirinktu lokalsios paieškos metodu: <http://www.mymathlib.com/optimization/nonlinear/unconstrained/>