

Lab Topic 5

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PSO codes



Clone the repository: **GitHub**
→ **SDMBIGDAT19** (Store it outside SandBox)



The codes and slides supplement the textbook



Lectures delivered at the BigDat19 5th International Winter school on Big Data, Cambridge University, UK (Jan, 2019)



<http://bigdat2019.irdta.eu/>

We will look at the following codes in **SDMBIGDAT19/CODES**:

- ▶ **r2ss.m**: Helper function; no need to look inside
- ▶ **r2sv.m**: Helper function; no need to look inside
- ▶ **s2rs.m**: Helper function; no need to look inside
- ▶ **s2rv.m**: Helper function; no need to look inside
- ▶ **crcbchkstdsrchrng.m**: Helper function; no need to look inside
- ▶ **crcbpso.m**: Main PSO code that can be applied to any fitness function
- ▶ **crcbpsotestfunc.m**: A benchmark fitness function; Also an example for how to code fitness functions to work with **crcbpso.m**
- ▶ **crcbqcfifunc.m**: The fitness function for quadratic chirp GLRT (in WGN)
- ▶ **crcbqcpso.m**: Applies PSO to the quadratic chirp fitness function
- ▶ **test_crcbpso.m**: Test function for **crcbpso.m**
- ▶ **test_crcbqcpso.m**: Test function for **crcbqcpso.m**

Exercise #1 Part 1

- ▶ Read the short user manual **CODES/CodeDoc.pdf**
- ▶ The main usage instructions are in the “help” for each function
- ▶ The **test_<funcName>.m** scripts show examples of usage for some of the functions
 - ▶ The **test_crcbpso.m** script shows how **crcbpso.m** is applied to a benchmark fitness function (defined in **crcbpsotestfunc.m**)

Exercise #1 Part 2

- ▶ Understand the concept of structures in Matlab
 - ▶ Matlab structures work in the same way as structures in C
 - ▶ `X = struct('a', 5.0, 'b', 6.0);`
 - ▶ `disp(X.a)` will show 5.0
 - ▶ `disp(X.b)` will show 6.0
- ▶ Structures offer a convenient way to move a large number of arguments into and out of a function
- ▶ Structures also help make your codes future-proof: New versions of codes can use new input arguments while old versions will ignore them

Exercise #1 Part 3

- ▶ Understand the concept of **function handle**
- ▶ A function handle is a variable that can be used to call a function
- ▶ $z=5.0$
- ▶ $F = \underbrace{@(x,y)}_{\substack{\text{What are the input} \\ \text{variables that will be} \\ \text{sent to } F?}} \quad \text{foo}(x,z,y)$
- ▶ F is a handle to function foo
- ▶ $F(2.0, 3.0)$ is the same as $\text{foo}(2.0, 5.0, 3.0)$
- ▶ The CRCBPSO code accepts as input the handle to the fitness function to be optimized
- ▶ This allows the same PSO code to be run on any fitness function
- ▶ Type “function handle” in Matlab’s “Search documentation” bar and read more about this feature

Anatomy of SDMBIGDAT19 codes

CRCBPSO: The PSO code

Inputs

- **Handle to fitness function**
 - Fitness function initialized with parameters that will not change from one call to another in PSO
 - Example: Data vector, PSD, sampling frequency
- **Number of search space dimensions** (= Number of parameters to maximize the likelihood Ratio over)
 - Example: 3 D search space for GLRT of quadratic chirp

Output

- **Structure** containing fields:
 - Best fitness value found
 - Location of the best fitness (in standardized coordinates)
 - Number of fitness evaluations (remember that particle fitness is not evaluated when they are outside the search space)

CRCBPSO: The PSO code

% S=CRCBPSO(Fhandle,N)

% Runs **local best PSO** on the fitness function with handle **Fhandle**. If Fname
% is the name of the function, **Fhandle = @(x) <Fname>(x, FP)**, where FP is
% the set of parameters needed by Fname.

Example: FP would be a structure that includes the data vector, PSD etc.

% N is the dimensionality of the
% fitness function.

%The output is returned in the **structure** S. The field
% of S are:

% 'bestLocation : Best location found (in standardized coordinates)

% 'bestFitness': Best fitness value found

% 'totalFuncEvals': Total number of fitness function evaluations.

Simple fitness function

```
F = CRCBPSOTESTFUNC(X,P)
```

Compute the Rastrigin fitness function for each row of X. X is standardized, that is $0 \leq X(i,j) \leq 1$.

Example of the matrix X for a 2D space

	Parameter 1	Parameter 2
Location 1	0.1	0.8
Location 2	0.5	0.2

The fitness values are returned in F.

Example of F for the above X

	Fitness
Location 1	10.0
Location 2	2.0

Simple fitness function

```
F = CRCBPSOTESTFUNC(X,P)
```

P has two arrays **P.rmin** and **P.rmax** that are used to convert $X(i,j)$ internally to actual coordinate values before computing fitness:

$$X(:,j) \rightarrow X(:,j) * (rmax(j) - rmin(j)) + rmin(j)$$

Example for 2D space: **P.rmin** = [-5,5]; **P.rmax** = [5, 10] \Rightarrow range for parameter 1 is [-5,5] and for parameter 2 is [5,10]

- The fields **rmin** and **rmax** are required for any fitness function
- In your fitness function, the structure **P** will need more fields. Example: **data**, **PSD**, etc.

Part where the input standardized coordinates are converted to real coordinates

```
%Check for out of bound coordinates and flag them
validPts = crcbchkstdsrchrng(xVec);
%Set fitness for invalid points to infty
fitVal(~validPts)=inf;
%Convert valid points to actual locations
xVec(validPts,:) = s2rv(xVec(validPts,:),params);
```

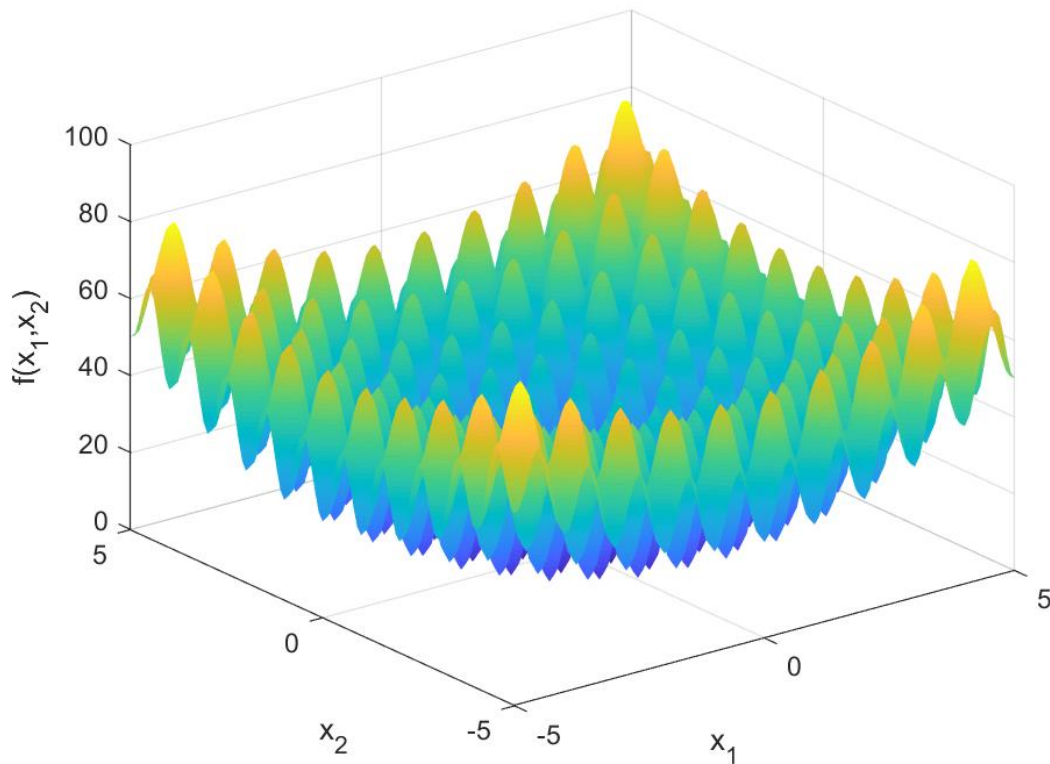
Simple fitness function

```
F = CRCBPSOTESTFUNC(X,P)
```

The main part of this function

```
for lpc = 1:nrows
    if validPts(lpc)
        % Only the body of this block should be replaced for different fitness
        % functions
        x = xVec(lpc,:);
        fitVal(lpc) = sum(x.^2-10*cos(2*pi*x)+10);
    end
end
```

Exercise #2



- ▶ Run **test_crcbpso.m**: it runs **crcbpso** on the standard benchmark fitness function **crcbpsotestfunc.m**
- ▶ Experiment:
 - ▶ Change number of dimensions
 - ▶ Change search range
- ▶ **Optional challenge**: Learn how to use the **surf** function in Matlab and make the plot shown (see Matlab documentation for **surf** and worked out examples)

Optional Exercise

- Code a matlab function for any one benchmark fitness function following the example of `crcbpsotestfunc.m`
 - Different teams can pick different functions
 - Ignore the fitness functions that have **red highlights** (there are typographical errors in them)
 - Column “D”: Dimensionality of search space
 - Column “Feasible Bounds”: Range of each coordinate defining the search space, which is a hypercube
- Apply `crcbpbso` to the fitness function and find the solution for the global minimum and minimizer

TABLE I
BENCHMARK FUNCTIONS

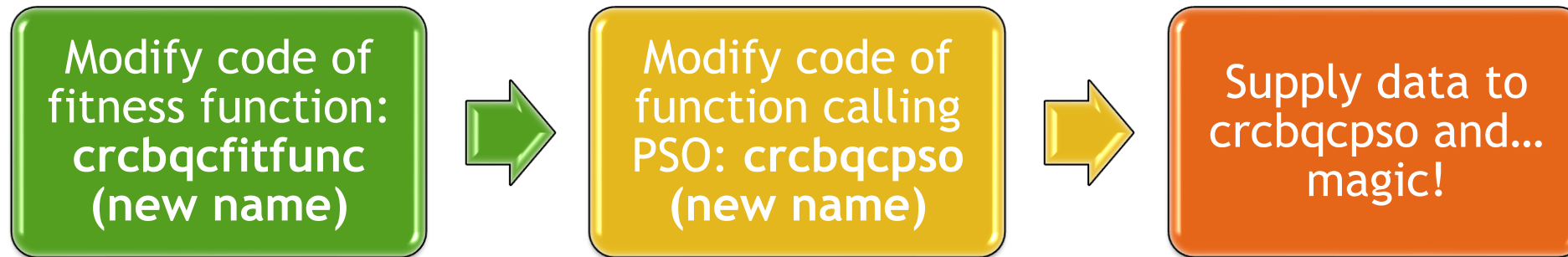
Equation	Name	D	Feasible Bounds
$f_1 = \sum_{i=1}^D x_i^2$	Sphere/Parabola	30	$(-100, 100)^D$
$f_2 = \sum_{i=1}^D (\sum_{j=1}^i x_j)^2$	Schwefel 1.2	30	$(-100, 100)^D$
$f_3 = \sum_{i=1}^{D-1} \{100 (x_{i+1} - x_i^2)^2 + (x_i - 1)^2\}$	Generalized Rosenbrock	30	$(-30, 30)^D$
$f_4 = -\sum_{i=1}^D x_i \sin(\sqrt{ x_i })$	Generalized Schwefel 2.6	30	$(-500, 500)^D$
$f_5 = \sum_{i=1}^D \{x_i^2 - 10 \cos(2\pi x_i) + 10\}$	Generalized Rastrigin	30	$(-5.12, 5.12)^D$
$f_6 = -20 \exp \left\{ -0.2 \sqrt{\frac{1}{D} \sum_{i=1}^D x_i^2} \right\} - \exp \left\{ \frac{1}{D} \sum_{i=1}^D \cos(2\pi x_i) \right\} + 20 + e$	Ackley	30	$(-32, 32)^D$
$f_7 = \frac{1}{4000} \sum_{i=1}^D x_i^2 - \prod_{i=1}^D \cos\left(\frac{x_i}{\sqrt{i}}\right) + 1$	Generalized Griewank	30	$(-600, 600)^D$
$f_8 = \frac{\pi}{D} \left\{ 10 \sin^2(\pi y_i) + \sum_{i=1}^{D-1} (y_i - 1)^2 \{1 + 10 \sin^2(\pi y_{i+1})\} + (y_D - 1)^2 \right\} + \sum_{i=1}^D \mu(x_i, 10, 100, 4)$ $y_i = 1 + \frac{1}{4} (x_i + 1)$ $\mu(x_i, a, k, m) = \begin{cases} k(x_i - a)^m & x_i > a \\ 0 & -a \leq x_i \leq a \\ k(-x_i - a)^m & x_i < -a \end{cases}$	Penalized Function P8	30	$(-50, 50)^D$
$f_9 = 0.1 \left\{ \sin^2(3\pi x_1) + \sum_{i=1}^{D-1} (x_i - 1)^2 \{1 + \sin^2(3\pi x_{i+1})\} + (x_D - 1)^2 \times \{1 + \sin^2(2\pi x_D)\} \right\} + \sum_{i=1}^D \mu(x_i, 5, 100, 4)$	Penalized Function P16	30	$(-50, 50)^D$
$f_{10} = 4x_1^2 - 2.1x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2^4$	Six-hump Camel-back	2	$(-5, 5)^D$
$f_{11} = \left\{ 1 + (x_1 + x_2 + 1)^2 (19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2) \right\} \times \left\{ 30 + (2x_1 - 3x_2)^2 (18 - 32x_1 + 12x_1^2 + 48x_2 - 36x_1x_2 + 27x_2^2) \right\}$	Goldstein-Price	2	$(-2, 2)^D$
$f_{12} = -\sum_{i=1}^5 \left\{ \sum_{j=1}^4 (x_j - a_{ij})^2 + c_i \right\}^{-1}$ $f_{13} = -\sum_{i=1}^7 \left\{ \sum_{j=1}^4 (x_j - a_{ij})^2 + c_i \right\}^{-1}$	Shekel 5	4	$(0, 10)^D$
	Shekel 7	4	$(0, 10)^D$

► Main Exercise

Using PSO to get GLRT for the Quadratic Chirp in
Colored Gaussian Noise

Plan

- ▶ Develop code for the fitness function
- ▶ Develop code for calling PSO on the fitness function with **multiple independent runs** and picking the results from the best run
- ▶ Create a data realization and use codes to obtain GLRT using PSO
- ▶ Display results



Fitness function

Preliminaries

- ▶ You have already coded the fitness function required for quadratic chirp in colored Gaussian noise in Lab Topic 4
- ▶ You have to now **cast it into a form that can be called by the PSO code**
- ▶ This is best done by taking the fitness function already coded for WGN and modifying it by replacing some parts

Fitness function

`F = CRCBQCFITFUNC<New Name>(X, P)`

Compute the fitness function (~~sum of squared residuals~~
~~function after maximization over the amplitude~~
~~parameter~~→ **To be replaced by log-likelihood ratio for colored noise**
maximized over amplitude) for data containing the quadratic
chirp signal at the parameter values in X.

The fitness values are returned in F.

X is standardized, that is $0 \leq X(i, j) \leq 1$.

The fields `P.rmin` and `P.rmax` are used to convert `X(i, j)`
internally before computing the fitness: `X(:, j) ->`
`X(:, j) * (rmax(j) - rmin(j)) + rmin(j)`. $\Rightarrow \theta_i = x_i * (b_i - a_i) + a_i$

Fitness function

$F = \text{CRCBQCFITFUNC}(\text{New Name})(X, P)$

The fields `P.dataY` and `P.dataX` are used to transport the data and its time stamps. The fields `P.dataXSq` and `P.dataXCb` contain the timestamps squared and cubed respectively. **You will need an extra field to supply the PSD for colored noise.**

`[F,R] = CRCBQCFITFUNC(X,P)` returns the quadratic chirp coefficients corresponding to the rows of `X` in `R`. **Converts X to R**

`[F,R,S] = CRCBQCFITFUNC(X,P)` Returns the quadratic chirp signals corresponding to the rows of `X` in `S`. **Converts R to S**

% Only the body of this block should be replaced for different fitness functions

```
fitVal(lpc) = ssrqc(x, params);
```

function ssrVal = ssrqc(x,params) ← Can be replaced with the fitness function from Lab Topic 4 Or...

%Generate normalized quadratic chirp

The signal is generated inside this function for speed (calling crcbgenqcsig.m would be slow because dataX.^2 and dataX.^3 will be recomputed in every call but they need to be computed only once)

```
phaseVec = x(1)*params.dataX + x(2)*params.dataXSq + x(3)*params.dataXCb;
```

```
qc = sin(2*pi*phaseVec);
```

```
qc = qc/norm(qc); ⇒ norm is the one defined for WGN → Replace by the norm for given PSD
```

%Compute fitness

```
ssrVal = -(params.dataY*qc')^2; ⇒  $-\langle \bar{y}, \bar{q}(\theta) \rangle^2$  for WGN → Replace by the inner product for given PSD
```

Calling PSO

CRCBQCPSO: Inputs

```
%O = CRCBQCPSO—<New Name>(I,P,N)
%I is the input struct with the fields given below.
% 'dataY': The data vector (a uniformly sampled time series).
% 'dataX': The time stamps of the data samples.
% 'dataXSq': dataX.^2
% 'dataXCb': dataX.^3
```

New fields needed for colored Gaussian noise:

```
'psd' : PSD values at positive DFT frequencies
'sampFreq': Sampling frequency
% 'rmin', 'rmax': The minimum and maximum values of the three parameters
%                  a1, a2, a3 in the candidate signal:
%                  a1*dataX+a2*dataXSq+a3*dataXCb
```

```
P is the PSO parameter
%struct. Setting P to [] will invoke default parameters (see
CRCBPSO).
```

```
% N is the number of independent PSO runs.
```

```
%The output is returned in the struct O.
```

CRCBQCPSO: Outputs

```
%The fields of O are:  
% 'allRunsOutput': An N element struct array containing results  
%                   from each PSO run. The fields of this struct are:  
%   'fitVal': The fitness value.  
%   'qcCoefs': The coefficients [a1, a2, a3].  
%   'estSig': The estimated signal.  
%   'totalFuncEvals': The total number of fitness evaluations.  
% 'bestRun': The best run.  
% 'bestFitness': best fitness from the best run.  
% 'bestSig' : The signal estimated in the best run.  
% 'bestQcCoefs' : [a1, a2, a3] found in the best run.
```


Signal from estimated parameters

```
61 %Prepare output
62 fitVal = zeros(1,nRuns);
63 for lpruns = 1:nRuns
64     fitVal(lpruns) = outStruct(lpruns).bestFitness;
65     outResults.allRunsOutput(lpruns).fitVal = fitVal(lpruns);
66     [~,qcCoefs] = fHandle(outStruct(lpruns).bestLocation);
67     outResults.allRunsOutput(lpruns).qcCoefs = qcCoefs;
68     estSig = crcbgenqcsig(inParams.dataX,1,qcCoefs);
69     estAmp = inParams.dataY*estSig(:);
70     estSig = estAmp*estSig;
71     outResults.allRunsOutput(lpruns).estSig = estSig;
72     outResults.allRunsOutput(lpruns).totalFuncEvals = outStruct(lpruns).totalFuncEvals;
73 end
```

Convert standardized
coordinates to real
coordinates

Replace!

When obtaining the maximum of the likelihood ratio **over amplitude**, we get the solution: $A = \langle \bar{y}, \bar{q}(\theta) \rangle$ (and substituting this into the likelihood ratio gives $\langle \bar{y}, \bar{q}(\theta) \rangle^2$)
 $\therefore \text{estSig}(\bar{s}(\theta)) \rightarrow \text{template}(\bar{q}(\theta))$ by normalization to SNR=1 $\rightarrow A = \langle \bar{y}, \bar{q}(\theta) \rangle \rightarrow \text{estSig} = A \times \bar{q}(\theta)$

Data realization

Data realization

- ▶ Number of samples: 512
- ▶ Sampling frequency: 512 Hz
- ▶ **Noise realization: Toy PSD (Note change of numbers)**
$$\text{noisePSD} = @(f) (f \geq 50 \ \& \ f \leq 100) \cdot (f - 50) \cdot (100 - f) / 625 + 1;$$
 - ▶ You will need to generate the PSD at positive DFT frequencies
- ▶ **Signal: Quadratic chirp with the following parameters**
 - ▶ SNR=10
 - ▶ $a_1 = 10, a_2 = 3, a_3 = 3$
- ▶ **Data: Noise realization + signal**

GLRT calculation

PSO search range

- ▶ Number of independent PSO runs = 8
- ▶ Number of iterations = 1000
- ▶ Search ranges
 - ▶ $a_1 \in [1, 180]$
 - ▶ $a_2 \in [1, 10]$
 - ▶ $a_3 \in [1, 10]$
- ▶ \Rightarrow We are searching over signals whose
 - ▶ **final instantaneous frequency** is in the range $[1 + 2 + 3, 180 + 20 + 30] = [6, 230]$ Hz
 - ▶ Initial instantaneous frequency is 1 Hz.
- ▶ This range is 89.4% of the widest possible range $[0, 256]$ Hz

Presentation of results

Presenting results

- Plot the data
- Plot the true signal on top
- Plot the best estimated signal on top
- (Optional) You can also try to plot the best signals from all the runs of PSO in a different color: This will show the **spread in performance** of PSO

