

# Projection onto Balls

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## Contents

<b>1</b>	<b>Projection onto the <math>L_1</math> Ball</b>	<b>2</b>
<b>2</b>	<b>Projection onto the <math>L_2</math> Ball</b>	<b>4</b>
<b>3</b>	<b>Projection onto the <math>L_\infty</math> Ball</b>	<b>6</b>
<b>4</b>	<b>Projection onto the Simplex</b>	<b>8</b>

# 1 Projection onto the $L_1$ Ball

$$\arg \min_{\|x\|_1 \leq r} \left\{ \frac{1}{2} \|x - y\|_2^2 \right\}$$

The Lagrangian of the problem can be written as:

$$\begin{aligned} L(x, \lambda) &= \frac{1}{2} \|x - y\|^2 + \lambda (\|x\|_1 - r) \\ &= \sum_{i=1}^n \left( \frac{1}{2} (x_i - y_i)^2 + \lambda |x_i| \right) - \lambda r \quad \text{Component wise form} \end{aligned}$$

The Dual Function is given by:

$$g(\lambda) = \inf_x L(x, \lambda)$$

The above can be solved component wise for the term  $\left( \frac{1}{2} (x_i - y_i)^2 + \lambda |x_i| \right)$  which is solved by the Soft Thresholding Operator:

$$x_i^* = \text{sign}(y_i) (|y_i| - \lambda)_+$$

Where  $(t)_+ = \max(t, 0)$ .

Clearly, when  $\lambda = 0$  it suggests that  $x = y$  and  $\|y\|_1 \leq r$ . For the case  $\lambda > 0$ , which suggests  $\|y\|_1 > 1$ , all needed is to find the optimal  $\lambda > 0$  which is given by the root of the objective function (Which is the constrain of the KKT System):

$$\begin{aligned} h(\lambda) &= \sum_{i=1}^n |x_i^*(\lambda)| - r \\ &= \sum_{i=1}^n (|y_i| - \lambda)_+ - r \end{aligned}$$

Namely, the solution is given by  $\lambda^* = \lambda : h(\lambda) = 0$ . The above is a Piece Wise linear function of  $\lambda$  and its Derivative given by:

$$\begin{aligned} \frac{d}{d\lambda} h(\lambda) &= \frac{d}{d\lambda} \sum_{i=1}^n (|y_i| - \lambda)_+ \\ &= \sum_{i=1}^n -\mathbf{1}_{\{|y_i| - \lambda > 0\}} \end{aligned}$$

Hence it can be solved using Newton Iteration.

Listing 1: MATLAB Code -  $L_1$  Ball Projection

```

1  function [ vX ] = ProjectL1Ball( vY, ballRadius, stopThr )
2  % ...
   %
3  % [ vX ] = ProjectL1Ball( vY, ballRadius, stopThr )
4  % Solving the Orthogonal Porjection Problem of the input ...
   vector onto the
5  % L1 Ball using Dual Function and Newtin Iteration.
6  % Input:
7  % - vY          - Input Vector.
8  %                Structure: Vector (Column).
9  %                Type: 'Single' / 'Double'.
10 %                Range: (-inf, inf).
11 % - ballRadius  - Ball Radius.
12 %                Sets the Radiuf of the L1 Ball. For ...
   Unit L1 Ball
13 %                set to 1.
14 %                Structure: Scalar.
15 %                Type: 'Single' / 'Double'.
16 %                Range: (0, inf).
17 % - stopThr     - Stopping Threshold.
18 %                Sets the trheold of the Newton ...
   Iteration. The
19 %                absolute value of the Objective ...
   Function will be
20 %                below the threshold.
21 %                Structure: Scalar.
22 %                Type: 'Single' / 'Double'.
23 %                Range: (0, inf).
24 % Output:
25 % - vX          - Output Vector.
26 %                The projection of the Input Vector ...
   onto the Simplex
27 %                Ball.
28 %                Structure: Vector (Column).
29 %                Type: 'Single' / 'Double'.
30 %                Range: (-inf, inf).
31 % References
32 % 1. https://math.stackexchange.com/questions/2327504.
33 % 2. https://en.wikipedia.org/wiki/Newton%27s\_method.
34 % Remarks:
35 % 1. a
36 % TODO:
37 % 1. U.
38 % Release Notes:
39 % - 1.0.001      29/06/2017 Royi Avital
40 % * Enforcing Lambda to be non negative (Dealing ...
   with the case 'vY'
41 % is obeying || vY ||_1 ≤ ballRadius).
42 % - 1.0.000      27/06/2017 Royi Avital
43 % * First release version.
44 % ...
   %
45
46 FALSE = 0;

```

```

47     TRUE      = 1;
48
49     OFF       = 0;
50     ON        = 1;
51
52     paramLambda = 0;
53     % The objective functions which its root (The 'paramLambda' ...
54     % which makes it
55     % vanish) is the solution
56     objVal      = sum(max(abs(vY) - paramLambda, 0)) - ...
57     ballRadius;
58
59     while(abs(objVal) > stopThr)
60         objVal      = sum(max(abs(vY) - paramLambda, 0)) - ...
61         ballRadius;
62         df          = sum(-(abs(vY) - paramLambda) > 0)); %<! ...
63         % Derivative of 'objVal' with respect to Lambda
64         paramLambda = paramLambda - (objVal / df); %<! Newton ...
65         % Iteration
66     end
67
68     % Enforcing paramLambda ≥ 0. Otherwise it suggests || vY ...
69     % || -1 ≤ ballRadius.
70     % Hence the Optimal vX is given by vX = vY.
71     paramLambda = max(paramLambda, 0);
72
73     vX = sign(vY) .* max(abs(vY) - paramLambda, 0);
74
75 end

```

## 2 Projection onto the $L_2$ Ball

$$\arg \min_{\|x\|_2 \leq r} \left\{ \frac{1}{2} \|x - y\|_2^2 \right\}$$

The Lagrangian is given by:

$$L(x, \lambda) = \frac{1}{2} \|x - y\|_2^2 + \lambda (x^T x - r)$$

The problem above is convex and Slater's condition holds by choosing  $x = \mathbf{0}$ .

The KKT Conditions are given by:

$$\begin{aligned} \nabla_x L(x, \lambda) &= x - y + 2\lambda x = 0 & (1) \text{ Stationary} \\ \lambda (x^T x - r) &= 0 & (2) \text{ Slackness} \\ x^T x - r &\leq 0 & (3) \text{ Primal Feasibility} \\ \lambda &\geq 0 & (4) \text{ Dual Feasibility} \end{aligned}$$

From (1) one could see that if  $\lambda = 0$  then  $x = y$  and from (3) it means  $x^T x = y^T y \leq r$ . If  $\lambda > 0$  then from (2) it means  $x^T x = r$ . Taking (1) and multiply it by  $x^T$  yields (Remember  $x^T x = r$ ):

$$x^T x - x^T y + 2\lambda x^T x = 0 \Rightarrow \lambda = \frac{x^T y - r}{2r}$$

Plugging the result back into (1) yields  $x = \left(1 + \frac{x^T y - r}{r}\right)^{-1} y = \frac{r}{x^T y} y$ . Since  $x^T x = r$  and  $x$  is a scaled version of  $y$  (Namely has the same direction as  $y$ ) which results in:

$$x = \begin{cases} y & \text{if } \|y\|_2 \leq r \\ r \frac{y}{\|y\|_2} & \text{if } \|y\|_2 > r \end{cases}$$

Listing 2: MATLAB Code -  $L_2$  Ball Projection

```

1  function [ vX ] = ProjectL2Ball( vY, ballRadius )
2  % ...
   % -----
3  % [ vX ] = ProjectL2Ball( vY, ballRadius, stopThr )
4  % Solving the Orthogonal Porjection Problem of the input ...
   % vector onto the
5  % L1 Ball.
6  % Input:
7  % - vY          - Input Vector.
8  %               Structure: Vector (Column).
9  %               Type: 'Single' / 'Double'.
10 %               Range: (-inf, inf).
11 % - ballRadius  - Ball Radius.
12 %               Sets the Radius of the L2 Ball. For ...
   % Unit L2 Ball
13 %               set to 1.
14 %               Structure: Scalar.
15 %               Type: 'Single' / 'Double'.
16 %               Range: (0, inf).
17 % Output:
18 % - vX          - Output Vector.
19 %               The projection of the Input Vector ...
   % onto the L2
20 %               Ball.
21 %               Structure: Vector (Column).
22 %               Type: 'Single' / 'Double'.
23 %               Range: (-inf, inf).
24 % References
25 % 1. h
26 % Remarks:
27 % 1. a
28 % TODO:
29 % 1. U.
30 % Release Notes:
31 % - 1.0.000      29/06/2017 Royi Avital
32 % * First release version.
```

```

33      % ...
34      %
35      FALSE = 0;
36      TRUE  = 1;
37
38      OFF   = 0;
39      ON    = 1;
40
41      vX = min((ballRadius / norm(vY, 2)), 1) * vY;
42
43
44      end

```

### 3 Projection onto the $L_\infty$ Ball

$$\arg \min_{\|x\|_\infty \leq r} \left\{ \frac{1}{2} \|x - y\|_2^2 \right\}$$

Since  $\|x\|_\infty = \max_i |x_i|$ ,  $i = 1, 2, \dots, n$  the above can be written as:

$$\begin{aligned} & \arg \min_x \quad \frac{1}{2} \|x - y\|_2^2 \\ & \text{subject to} \quad x_i \leq r \quad i = 1, 2, \dots, n \\ & \quad \quad \quad -x_i \leq r \quad i = 1, 2, \dots, n \end{aligned}$$

The Lagrangian is given by:

$$L(x, \lambda_1, \lambda_2) = \frac{1}{2} \|x - y\|_2^2 + \lambda_1^T (x - r) + \lambda_2^T (-x - r)$$

The problem above is convex and Slater's condition holds by choosing  $x = \mathbf{0}$ .

The KKT Conditions are given by:

$$\begin{aligned} \nabla_x L(x, \lambda_1, \lambda_2) &= x - y + \lambda_1 x - \lambda_2 x = 0 & (1) \text{ Stationary} \\ \lambda_{1,i} (x_i - r) &= 0 & (2) \text{ Slackness} \\ \lambda_{2,i} (-x_i - r) &= 0 & (3) \text{ Slackness} \\ x_i - r &\leq 0 & (4) \text{ Primal Feasibility} \\ -x_i - r &\leq 0 & (5) \text{ Primal Feasibility} \\ \lambda_{1,i}, \lambda_{2,i} &\geq 0 & (6) \text{ Dual Feasibility} \end{aligned}$$

As can be seen from above, the problem can be solved component wise. Hence the sub script  $i$  for the Lagrange Multiplier will be neglected. From

(2) and (3) it can be shown that either  $\lambda_1 > 0$  or  $\lambda_2 > 0$  but not both. Hence if  $\lambda_1 = 0$  then  $\lambda_2 = 0$  which means  $x_i = y_i$  and  $|x_i| \leq 1$ . Moreover, if  $\lambda_1 > 0$  then  $\lambda_2 = 0$  hence  $x_i = r$  and  $y_i > 1$ . The same goes the other way around which yields:

$$x_i = \text{sign}(y_i) \min\{r, |y_i|\}, \quad i = 1, 2, \dots, n$$

Listing 3: MATLAB Code -  $L_\infty$  Ball Projection

```

1  function [ vX ] = ProjectLInfBall( vY, ballRadius )
2  % ...
   %
3  % [ vX ] = ProjectL2Ball( vY, ballRadius, stopThr )
4  % Solving the Orthogonal Porjection Problem of the input ...
   vector onto the
5  % L Inf Ball.
6  % Input:
7  % - vY          - Input Vector.
8  %               Structure: Vector (Column).
9  %               Type: 'Single' / 'Double'.
10 %               Range: (-inf, inf).
11 % - ballRadius  - Ball Radius.
12 %               Sets the Radius of the L Inf Ball. ...
   For Unit L Inf
13 %               Ball set to 1.
14 %               Structure: Scalar.
15 %               Type: 'Single' / 'Double'.
16 %               Range: (0, inf).
17 % Output:
18 % - vX          - Output Vector.
19 %               The projection of the Input Vector ...
   onto the L Inf
20 %               Ball.
21 %               Structure: Vector (Column).
22 %               Type: 'Single' / 'Double'.
23 %               Range: (-inf, inf).
24 % References
25 % 1. h
26 % Remarks:
27 % 1. a
28 % TODO:
29 % 1. U.
30 % Release Notes:
31 % - 1.0.000      29/06/2017 Royi Avital
32 % * First release version.
33 % ...
   %
34
35 FALSE = 0;
36 TRUE  = 1;
37
38 OFF   = 0;
39 ON    = 1;

```

```

40
41     vX = sign(vY) .* min(abs(vY), ballRadius);
42
43
44     end

```

## 4 Projection onto the Simplex

$$\arg \min_{x \succeq 0, \mathbf{1}^T x = r} \left\{ \frac{1}{2} \|x - y\|_2^2 \right\}$$

The Lagrangian of the problem can be written as (Leaving the non negativity constrain implicit):

$$L(x, \mu) = \frac{1}{2} \|x - y\|_2^2 + \mu (\mathbf{1}^T x - r)$$

The Dual Function is given by (Includes the Non Negativity Constrain):

$$\begin{aligned}
g(\mu) &= \inf_{x \succeq 0} L(x, \mu) \\
&= \inf_{x \succeq 0} \sum_{i=1}^n \left( \frac{1}{2} (x_i - y_i)^2 + \mu x_i \right) - \mu r && \text{Component wise form} \\
&= \inf_{x \succeq 0} \frac{1}{2} \sum_{i=1}^n (x_i - (y_i - \mu))^2 + \mu (\mathbf{1}^T y - r) + n\mu^2
\end{aligned}$$

The minimization with respect to  $x_i$  is basically a projection problem into  $\mathbb{R}_+$  of  $y_i - \mu$  which is given by:

$$x_i^* = (y_i - \mu)_+$$

The solution is given by finding the  $\mu$  which holds the equality constrain. In the  $L_1$  case above the constrain was in inequality form hence  $\lambda$  had to be non negative Yet the above is equality constrain hence  $\mu$  can have any value and it is not limited to non negativity as  $\lambda$  in the  $L_1$  case.

The function, From the KKT, which enforces equality, is given by:

$$h(\mu) = \sum_{i=1}^n x_i^* - r = \sum_{i=1}^n (y_i - \mu)_+ - r$$



Namely, the solution is given by  $\mu^* = \mu : h(\mu) = 0$ . The above is a Piece Wise linear function of  $\mu$  and its Derivative given by:

$$\begin{aligned}\frac{d}{d\mu}h(\mu) &= \frac{d}{d\mu} \sum_{i=1}^n (y_i - \mu)_+ \\ &= \sum_{i=1}^n -\mathbf{1}_{\{y_i - \mu > 0\}}\end{aligned}$$

Hence it can be solved using Newton Iteration.

Listing 4: MATLAB Code - Simplex Projection

```

1  function [ vX ] = ProjectSimplex( vY, ballRadius, stopThr )
2  % ...
   %
3  % [ vX ] = ProjectSimplex( vY, ballRadius, stopThr )
4  % Solving the Orthogonal Porjection Problem of the input ...
   % vector onto the
5  % Simplex Ball using Dual Function and Newton Iteration.
6  % Input:
7  % - vY          - Input Vector.
8  %               Structure: Vector (Column).
9  %               Type: 'Single' / 'Double'.
10 %               Range: (-inf, inf).
11 % - ballRadius  - Ball Radius.
12 %               Sets the Radius of the Simplex Ball. ...
   % For Unit
13 %               Simplex set to 1.
14 %               Structure: Scalar.
15 %               Type: 'Single' / 'Double'.
16 %               Range: (0, inf).
17 % - stopThr     - Stopping Threshold.
18 %               Sets the trheolds of the Newton ...
   % Iteration. The
19 %               absolute value of the Objective ...
   % Function will be
20 %               below the threshold.
21 %               Structure: Scalar.
22 %               Type: 'Single' / 'Double'.
23 %               Range: (0, inf).
24 % Output:
25 % - vX          - Output Vector.
26 %               The projection of the Input Vector ...
   % onto the Simplex
27 %               Ball.
28 %               Structure: Vector (Column).
29 %               Type: 'Single' / 'Double'.
30 %               Range: (-inf, inf).
31 % References
32 % 1. https://math.stackexchange.com/questions/2327504.
33 % 2. https://en.wikipedia.org/wiki/Newton%27s\_method.
34 % Remarks:

```

```

35 % 1. a
36 % TODO:
37 % 1. U.
38 % Release Notes:
39 % - 1.0.001 09/05/2017 Royi Avital
40 % * Renaming 'paramLambda' -> 'paramMu' to match ...
    derivation.
41 % - 1.0.000 09/05/2017 Royi Avital
42 % * First release version.
43 % ...
    -----
    %
44
45 FALSE = 0;
46 TRUE  = 1;
47
48 OFF   = 0;
49 ON    = 1;
50
51 % Choosing paramMu = min(vY) - ballRadius yields starting ...
    value of paramMu
52 % which ensures the objective value to have positive value ...
    -> Easier to
53 % find its root.
54 paramMu = min(vY) - ballRadius;
55 % The objective functions which its root (The 'paramMu' ...
    which makes it
56 % vanish) is the solution
57 objFun   = sum( max(vY - paramMu, 0) ) - ballRadius;
58
59 while(abs(objFun) > stopThr)
60     objFun   = sum( max(vY - paramMu, 0) ) - ballRadius;
61     df       = sum(-(vY - paramMu) > 0)); %<! Derivative ...
        of 'objVal' with respect to Mu
62     paramMu  = paramMu - (objFun / df); %<! Newton Iteration
63 end
64
65 vX = max(vY - paramMu, 0);
66
67
68 end

```