MATLAB code for CLAD with MIP model

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CladCompute.m

```
function [value,estimates,time,quality]=CladCompute()
%Main Program: Computes Max score by defining a MILP and calling milp.m
tic
[X,y,w] = readXyw();
[X,mu,sigma] = standardizeX(X,y);
[c,A,b] = definecAb(X,y,w);
[lb,ub, Aeq, beq, n, p, best] = definel bub(X,y);
%[x,cost,feasible]=milp(c,A,b,Aeq,beq,lb,ub,n, best);
 [x,cost,feasible, time]=milp_cplex(c,A,b,Aeq,beq,lb,ub,n, p, best);
estimatesNorm=x
value=cost
status=feasible
runtime=time
estimatesRaw=denormalizeEstimates(estimatesNorm,mu,sigma)
estimates=estimatesRaw
%estimatesRaw=denormalizeEstimates(estimatesNorm,mu,sigma)
estimates=estimatesRaw
value=cost
time=toc
quality=feasible;
end
```

definecAb.m

```
function [c,A,b]=definecAb(X,y,w)
%Defines c,A,b for milp.m

n=size(X,1);
p=size(X,2);

c3=repmat(0,1,n); %gammas integer
c1=repmat(0,1,p); %betas
c2=repmat(0,1,n); %phis
c4=repmat(1,1,n); %sm's
c5=repmat(1,1,n); %sp's

c=[c3 c1 c2 c4 c5];

d=20;
```

```
%d=5;
for i=1:1:n
   M(i)=abs(X(i,1))*d+abs(X(i,2:end))*repmat(d,1,p-1)';
end
for i=1:n
                %constraint 3c in pdf ssrn
  A1(i,1:n)=0; %gammas
  for j=1:p
  A1(i,(n+j))=X(i,j); %betas
  A1(i,(n+p+1):(2*n+p))=-1.*(1:n==i); %phis
  A1(i,(2*n+p+1):(3*n+p))=0;
                                       %sm's
  A1(i,(3*n+p+1):(4*n+p))=0;
                                       %sp's
end
for i=1:n
  b1(i)=0;
end
for i=1:n
              %constraint 3e in pdf ssrn
  A2(i,1:n)=M(i).*(1:n==i); %gammas
  for j=1:p
  A2(i,(n+j))=-X(i,j); %betas
  A2(i,(n+p+1):(2*n+p))=1.*(1:n==i);
                                         %phis
  A2(i,(2*n+p+1):(3*n+p))=0;
                                 %sm's
  A2(i,(3*n+p+1):(4*n+p))=0;
                                  %sp's
end
for i=1:n
  b2(i)=M(i);
end
for i=1:n
              %constraint 3f in pdf ssrn
  A3(i,1:n)=-M(i).*(1:n==i); %gammas
  A3(i,(n+1):(n+p))=0; %betas
  A3(i,(n+p+1):(2*n+p))=1.*(1:n==i); %phis
  A3(i,(2*n+p+1):(3*n+p))=0; %sm's
  A3(i,(3*n+p+1):(4*n+p))=0; %sp's
end
for i=1:n
  b3(i)=0; % for left censoring at 0
end
A = [A1 ; A2 ; A3];
b= [b1 b2 b3]';
end
```

```
function [lb,ub, Aeq, beq, n, p, best] = definelbub(X,y)
%Defines lb,ub for milp.m
d=20;
%d=10;
%d=5;
n=size(X,1);
p=size(X,2);
lb1=repmat(0,1,n);
                     %gammas
lb2=repmat(-d,1,p);
                      %betas
%lb3=repmat(-Inf,1,n); %phis
                         %phis, left censoring at zero (0)
lb3=repmat(0,1,n);
lb4=repmat(0,1,n);
                         %sm's
lb5=repmat(0,1,n);
                          %sp's
lb=[lb1 lb2 lb3 lb4 lb5];
BIGNUM=+Inf;
                     %gammas
ub1=repmat(1,1,n);
ub2=repmat(d,1,p);
                      %betas
ub3=repmat(+BIGNUM,1,n); %phis
ub4=repmat(+BIGNUM,1,n); %sm's
ub5=repmat(+BIGNUM,1,n);
                           %sp's
ub=[ub1 ub2 ub3 ub4 ub5];
for i=1:n
  Aeq(i,1:n)=0; %gammas
  Aeq(i,(n+1):(n+p))=0; %betas
  Aeq(i,(n+p+1):(2*n+p))=-1.*(1:n==i); %phis
  Aeq(i,(2*n+p+1):(3*n+p))=1.*(1:n==i); %sm's
  Aeq(i,(3*n+p+1):(4*n+p)) = -1.*(1:n==i); %sp's
end
for i=1:n
  beq(i) = -y(i);
end
best=+Inf;
end
```

```
function [x,cost,feasible]=milp(c,A,b,Aeq,beq,lb,ub,n, best);
% Solves a integer lp using branch and bound proceeding in a
% depth-first manner (recursive)
% c: is cost
                 A: is constraint matrix b: is constraint vector
% lb: lower bound ub: upper bound n: number of integer variables
% best: is best solution so far
% Note this traverses along the right most branch
% unlike the usual way of checking both branches before continuing. Also,
% it assumes first n variables must be integer.
% Written by Kevin Tomsovic
% Note: Last accessed, Autumn 2007, from K. Tomsovic's homepage.
% Note: by K.Florios, Date: 4.4.2016
% Initialization
feasible=0;
                % 1 - Feasible, 0 - Infeasible
                 % Tolerance on integers
tol=1e-3;
%%%tol=1e-6;
                    % Tolerance on integers
cost=inf;
% Solve current linprog, suppress error messages
options=optimset('display','off');
[x,cost,how]=linprog(c,A,b,Aeq,beq,lb,ub,[],options);
% If infeasible or cost is greater than bound (best solution found)
  then stop and return with infinite cost
if how<=0 | cost>best, cost=inf; return; end; %KJF comments out 1.4.2016
% Take first non integer u and branch
branch=0;
for i=1:n % Split LP on first non-integer u
  if abs(x(i)-round(x(i)))>tol, branch=i; break; end;
end;
% If integer solution then return optimal solution
if branch==0, feasible=1; % If valid integer solution (branch==0), return
  return;
else % Solve subproblems by putting new upper and lower bounds on branch variable
     % Proceeding depth-first
  lba=lb; uba=ub;
  uba(branch)=floor(x(branch)); lba(branch)=ceil(x(branch));
  % Solve left branch (x<=floor(x))
  [x,cost,feasible]=milp(c,A,b,Aeq,beq,lb,uba,n,best);
  % Solve right branch (x>=ceil(x))with new best
  best=min(cost,best);
  [xb,cost,how]=milp(c,A,b,Aeq,beq,lba,ub,n,best);
  if cost<best, x=xb; % Better solution found</pre>
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```
else cost=best;
end;
feasible=feasible | how; % Feasible solution found?
end;
return;
```

milp_cplex.m

```
function [x,deviation,feasible, time]=milp_cplex(c,A,b,Aeq,beq,lb,ub,n, p, best);
% Solves a mixed integer lp using gurobi 5.0.1
% c: is objective function coefficients A: is constraint matrix b: is constraint
   vector
% lb: lower bound ub: upper bound n: number of 0-1 variables
% best: is best solution so far
% Note this uses the MATLAB/Gurobi Interface documented at
% http://www.gurobi.com/documentation/5.0/reference-manual/node650
% Also, it assumes first n variables must be integer.
% The MIP equations of the maximum score estimator are available at
% Florios. K, Skouras, S. (2008) Exact computation of maximum weighted
% score estimators, Journal of Econometrics 146, 86-91.
% Written by Kostas Florios, July 20, 2012
% gurobi 5.0.1
% cd c:/Users/jones/gurobi500/win64/matlab
% gurobi_setup
%
% cd C:\gurobi501\win32\matlab
% gurobi_setup
% cplex 12.6
% addpath('C:\Program
   Files\IBM\ILOG\CPLEX_Studio_Preview126\cplex\matlab\x86_win32')
% Greg corei3 2016
% addpath('D:\Program
   Files\IBM\ILOG\CPLEX_Studio_Preview126\cplex\matlab\x86_win32')
model.Aineq = sparse(A) ;
model.f = c ;
model.bineq = b ;
model.Aeq = sparse(Aeq) ;
model.beq = beq';
model.lb = lb ;
model.ub = ub ;
%model.ctype = [repmat('B',size(b,2),1) ; repmat('C',size(c,2)-size(b,2),1)]' ;
model.ctype = [repmat('B',n,1) ; repmat('C',p,1) ; repmat('C',n,1) ;
   repmat('C',n,1); repmat('C',n,1)]';
opt = cplexoptimset('cplex') ;
```

```
opt.mip.display = 4 ;
opt.mip.interval = 1000 ;
opt.timelimit= 1800 ;
opt.mip.tolerances.mipgap = 0.00 ;
opt.parallel = 1 ;
opt.threads = 1 ;
opt.mip.strategy.file = 3 ;
opt.workmem = 1024;
opt.emphasis.mip = 3 ;
opt.exportmodel =
   'G:\KOSTAS\Core-i5-Laptop\matlab2011\milp-cplex-2016-for-CLAD\myModel.lp';
model.options = opt;
[x,fval,exitflag,output] = cplexmilp(model) ;
 fprintf('Optimization returned status: %s\n', output.message);
 fprintf('Objective Value: %e\n', fval);
 fprintf('(Wall clock) Time elapsed (s): %e\n', output.time);
 fprintf('Decision variables: show only the betas\n');
 %disp(x)
x=x((n+1):(n+p))
deviation=fval;
feasible=output.message;
time=output.time;
return;
   readXyw.m
```

```
function [X,y,w]=readXyw()

%Reads X,y,w of given max score problem

X=load('X_200_4.txt');
%X=[X(:,2) X(:,3) X(:,4) X(:,5)];
X=X(:,2:end);
y=load('ys_200_4.txt');
y=[y(:,2)];
%w=load('w_numeric5.txt');
%w=[w(:,2)];
w=repmat(1,size(X,1),1);
end
```

mu=mean(X);
sigma=std(X);

```
function [estimatesRaw] = denormalizeEstimates(estimatesNorm, mu, sigma);
%denormalized estimatesNorm obtained by Gurobi MIP to estimatesRaw, which
%are meaningful to the user
%quick and dirty implementation, based on GAMS and Fortran Analogues
p=size(estimatesNorm,1);
betaNorm=estimatesNorm;
for j=1:1:p
   betaRaw(j)=0;
   betaHelp(j)=0;
end
for j=1:1:p
   if (sigma(j) ~=0)
       betaHelp(j)= betaNorm(j)./sigma(j);
   end
   if (sigma(j) == 0)
       for jj=1:1:p
           if (sigma(jj) ~=0)
              betaHelp(j)=betaHelp(j)-betaNorm(jj).*mu(jj)./sigma(jj) ;
           else
              jj0=jj;
           end
       end
       betaHelp(j)=betaHelp(j)+betaNorm(jj0);
   end
end
for j=1:1:p
  %%betaRaw(j)=betaHelp(j)./betaHelp(1);
  betaRaw(j)=betaHelp(j);
end
estimatesRaw=betaRaw;
end
   standardizeX.m
function [X,mu,sigma] = standardizeX(X,y)
%Standardizes X
```

```
testX=(X-repmat(mu,size(X,1),1)) ./ repmat(sigma,size(X,1),1);
p=size(X,2);
for j=1:1:p
if isnan(testX(:,j))
    X(:,j) = X(:,j)
else
    X(:,j) = testX(:,j)
end
end
```

1 Contained functions

- 1. CladCompute.m \rightarrow main
- 2. **definecAb.m** \rightarrow defines matrices c,A,b
- 3. **definelbub.m** \rightarrow defines bounds lb,ub and matrices Aeq,beq
- 4. denormalizeEstimates.m \rightarrow post-processing
- 5. milp.m \rightarrow calls linprog
- 6. $milp_cplex.m \rightarrow calls$ cplexmilp commercial function
- 7. $\mathbf{readXyw.m} \rightarrow \mathbf{input}$ from files X.txt and ys.txt
- 8. standardizeX.m \rightarrow pre-processing

2 Guidelines for usage

We present the source code in Matlab in order to compute the CLAD estimator exactly (when technically possible) using MIP following (Bilias, Florios, & Skouras, 2013). There are several options on which MIP solver to use in MATLAB. Here are two options - for MATLAB (R) versions before R2014a:

- the milp.m solver by Prof. Kevin Tomsovic, for small scale examples ($N \le 30$), which uses the linprog() command for the LP subproblems solution
- the cplexmilp.m solver by IBM ILOG CPLEX OPTIMIZATION STUDIO 12.6 R which is formatted in the milp_cplex() function we wrote, and is the state-of-the-art solver for $N \approx 500$

For users of versions after R2014a (including R2014a):

• the intlinprog() solver by MATHWORKS ®, can be used for medium scale problems (N=200-400 observations)

Typical values for p are p \in {2,3,4,5,6}. Often CPLEX can tackle problems up to N \approx 500 and p < 5. 1 .

¹It is possible to choose the DGPs in such a way that problems with N ≈ 800 and p = 5 can be solved

References

Bilias, Y., Florios, K., & Skouras, S. (2013). Exact computation of censored least absolute deviations estimators (Technical Report). Athens University of Economics and Business. (Available at SSRN: http://ssrn.com/abstract=2372588)