The Max-kCut Problem

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Similar to the Max-Cut problem, the Max-kCut problem asks, given a graph $\mathbf{G} = (\mathbf{V}, \mathbf{E})$ and an integer k, does a cut exist of at least size k. For a given (weighted) adjacency matrix \mathbf{B} and integer k, the Max-kCut problem is formulated as the following primal problem

Here, $\mathbf{C} = -(1-1/k)/2 * (diag(\mathbf{B1}) - \mathbf{B})$. The Max-kCut problem is slightly more complex than the Max-Cut problem due to the inequality constraint. In order to turn this into a standard SQLP, we must replace the inequality constraints with equality constraints, which we do by introducing a slack variable \mathbf{x}^{l} , allowing the problem to be restated as

The function maxkcut takes as input an adjacency matrix B and an integer k, and returns the optimal solution using sqlp.

R> out <- maxkcut(B,k)

Numerical Example

To demonstrate the output provided by sqlp, consider the adjacency matrix

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R> data(Bmaxkcut) R> Bmaxcut V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 [1] 0 0 0 1 0 0 1 1 0

Lエ , 」	0	0	0	±	0	0	±	±	v	0
[2,]	0	0	0	1	0	0	1	0	1	1
[3,]	0	0	0	0	0	0	0	1	0	0
[4,]	1	1	0	0	0	0	0	1	0	1
[5,]	0	0	0	0	0	0	1	1	1	1

[6,]	0	0	0	0	0	0	0	0	1	0
[7,]	1	1	0	0	1	0	0	1	1	1
[8,]	1	0	1	1	1	0	1	0	0	0
[9,]	0	1	0	0	1	1	1	0	0	1
[10,]	0	1	0	1	1	0	1	0	1	0

Like the max-cut problem, here we are interested in the primal objective function. Like the max-cut problem, we take the negated value. We will use a value of k = 5 in the example.

R> out <- maxkcut(Bmaxkcut,5)

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R> -out$pobj
[1] 19
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Note also that the returned matrix ${\tt X}$ is a correlation matrix

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]
V1	1.000	0.381	0.503	-0.250	0.403	0.347	-0.250	-0.250	0.060	0.181
V2	0.381	1.000	0.231	-0.250	0.627	0.380	-0.250	0.160	-0.250	-0.250
VЗ	0.503	0.231	1.000	0.395	0.387	0.597	0.185	-0.250	0.074	0.089
V4	-0.250	-0.250	0.395	1.000	0.134	0.261	0.449	-0.250	0.163	-0.250
V5	0.403	0.627	0.387	0.134	1.000	0.348	-0.250	-0.250	-0.250	-0.250
V6	0.347	0.380	0.597	0.261	0.348	1.000	0.224	0.180	-0.250	0.239
V7	-0.250	-0.250	0.185	0.449	-0.250	0.224	1.000	-0.250	-0.250	-0.250
V8	-0.250	0.160	-0.250	-0.250	-0.250	0.180	-0.250	1.000	0.118	0.216
V9	0.060	-0.250	0.074	0.163	-0.250	-0.250	-0.250	0.118	1.000	-0.250
V10	0.181	-0.250	0.089	-0.250	-0.250	0.239	-0.250	0.216	-0.250	1.000