

## Appendix – Regime and Sociology

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## I. Coding solution for asymmetrical directional expectations

In the current software package 'QCA' for R—as in any other, as far as I know—there is no way of coding different directional expectations for the presence and absence of a condition. When calculating the intermediate solution, an expectation can be defined for each condition as either '1', '0', or '-' (don't care). Assigning `dir.exp=1` to a condition, automatically assumes the expectation '0' for the absences of the condition, and vice versa. Assigning '-' automatically assumes no expectations for both the presence and the absence of the conditions.

If one has asymmetrical directional expectations, this can lead to unwanted results. For example, if I would assign `dir.exp=1` for the condition THAW for the minimization of OUT, `~THAW` would be considered a difficult counterfactual and thus not be included in the minimization procedure for the intermediate solution term. However, because I do not have any expectations for `~THAW`, it could be included. In this case, I therefore assigned a directional expectation "-" (don't care), because then both THAW and `~THAW` can be included in the minimization of OUT. For the minimization of `~OUT`, I would want THAW to be excluded and `~THAW` included. Therefore I assigned a directional expectation of "0".

For the conditions NEIGH and CATH, the same reasoning applies, only in the reverse way. Here, I want NEIGH and CATH to be excluded from the minimization of OUT, but `~NEIGH` and `~CATH` should be included. This is achieved by assigning a directional expectation of "0". For the minimization of `~OUT`, I want both conditions in their presence and absence included. Therefore I assigned "-" to both.

The first table here below shows my directional expectations as formulated and explained in the manuscript (left column) and translates these into the decisions whether to include them in the minimization (right column). The second table shows how I coded the directional expectations in the calculation of the intermediate solutions in such a way that the same decisions on inclusion/exclusion follow.

This coding rule can also be expressed in words:

- Where the presence of the condition is expected to lead to the outcome and the absence of the condition has no attached expectation, a 'don't care'-expectation ('-') shall be assigned;
- Where the presence of the condition is expected to lead to the absence of outcome and the absence of the condition has no attached expectation, a negative expectation ('0') shall be assigned.

### Asymmetrical expectations and intended decisions

		Decision	
OUT	out	OUT	out

THAW	1	1	0	incl	excl
	0	-	-	incl	incl
NEIGH	1	0	1	excl	incl
	0	-	-	incl	incl
CATH	1	0	1	excl	incl
	0	-	-	incl	incl

### Symmetrical coding leading to the same intended decisions

			Decision	
	OUT	out	OUT	out
THAW	1	-	0	incl excl
	0	-	1	incl incl
NEIGH	1	0	-	excl incl
	0	1	-	incl incl
CATH	1	0	-	excl incl
	0	1	-	incl incl

## II. R Script

### 1. Loading the package and the data

```
> library(QCA)
```

To cite this package in publications, please use:

Dusa, Adrian (2019) QCA with R. A Comprehensive Resource.  
Springer International Publishing.

```
> DS <- read.csv("DS2019.csv")
> DS
  Country CON LIBDEM THAW NEIGH CATH OUT
1      AL    0     0    0     0    0    0
2      AT    0     1    0     1    1    0
3      BE    1     1    0     1    1    0
4      BG    0     0    0     0    0    0
5      CSSR  0     0    1     0    0    1
6      DK    0     1    0     0    0    1
7      FRG   1     1    0     0    0    1
8      FI    0     1    0     0    0    1
9      FR    1     1    0     0    0    1
10     GDR   0     0    0     0    0    0
11     GR    0     0    0     0    0    0
12     HU    0     0    0     0    0    0
13     IE    0     1    0     1    1    0
14     IT    1     1    0     0    1    0
15     NL    1     1    0     0    0    1
16     NO    0     1    0     0    0    1
17     PL    1     0    1     0    0    1
18     PT    0     0    0     0    1    0
19     RO    1     0    0     0    0    0
```

20	USSR	0	0	0	0	0	0
21	ES	0	0	0	0	1	0
22	SE	0	1	0	0	0	1
23	CH	0	1	0	1	0	0
24	GB	1	1	0	0	0	1
25	YU	0	0	1	0	0	1

## 2. Necessity analysis

### 2.1 Necessity analysis for the outcome

```
> superSubset(DS, outcome = "OUT", conditions = "CON, LIBDEM, THAW,
+           NEIGH, CATH", relation = "nec", incl.cut = 0.9, cov.cut = 0,
use.tilde
+           =
+           FALSE,
+           use.letters = FALSE)
```

		inclN	RoN	covN
1	neigh	1.000	0.286	0.524
2	cath	1.000	0.429	0.579
3	neigh*cath	1.000	0.500	0.611
4	con+LIBDEM	0.909	0.133	0.435
5	con+thaw	0.909	0.067	0.417
6	libdem+thaw	1.000	0.000	0.440
7	LIBDEM+THAW	1.000	0.643	0.688

### 2.2 Necessity analysis for the negated outcome

```
> > superSubset(DS, outcome = "~OUT", conditions = "CON, LIBDEM, THAW,
+           NEIGH, CATH", relation = "nec", incl.cut = 0.9, cov.cut = 0,
use.tilde
+           =
+           FALSE,
+           use.letters = FALSE)
```

		inclN	RoN	covN
1	thaw	1.000	0.273	0.636
2	con+LIBDEM	0.929	0.167	0.565
3	con+neigh	0.929	0.083	0.542
4	con+CATH	0.929	0.500	0.684
5	libdem+NEIGH	0.929	0.750	0.812
6	LIBDEM+neigh	1.000	0.000	0.560
7	libdem+CATH	0.929	0.750	0.812
8	neigh+CATH	0.929	0.083	0.542
9	con+NEIGH+cath	0.929	0.083	0.542

### 3. Sufficiency analysis

#### 3.1 Truth table

##### 3.1.1 Constructing the truth table for the outcome

*Note that row numbers in the output produced in R do not correspond with the row numbers in the manuscript.*

```
> ttDS2019 <- truthTable(DS, outcome = "OUT", conditions = "CON,  
LIBDEM, THAW, NEIGH, CATH", incl.cut = 0.7, complete = TRUE, show.cases  
= TRUE, sort.by = "out, n")  
> ttDS2019
```

OUT: output value  
n: number of cases in configuration  
incl: sufficiency inclusion score  
PRI: proportional reduction in inconsistency

	CON	LIBDEM	THAW	NEIGH	CATH	OUT	n	incl	PRI	cases
9	0	1	0	0	0	1	4	1.000	1.000	6,8,16,22
25	1	1	0	0	0	1	4	1.000	1.000	7,9,15,24
5	0	0	1	0	0	1	2	1.000	1.000	5,25
21	1	0	1	0	0	1	1	1.000	1.000	17
1	0	0	0	0	0	0	6	0.000	0.000	1,4,10,11,12,20
2	0	0	0	0	1	0	2	0.000	0.000	18,21
12	0	1	0	1	1	0	2	0.000	0.000	2,13
11	0	1	0	1	0	0	1	0.000	0.000	23
17	1	0	0	0	0	0	1	0.000	0.000	19
26	1	1	0	0	1	0	1	0.000	0.000	14
28	1	1	0	1	1	0	1	0.000	0.000	3
3	0	0	0	1	0	?	0	-	-	
4	0	0	0	1	1	?	0	-	-	
6	0	0	1	0	1	?	0	-	-	
7	0	0	1	1	0	?	0	-	-	
8	0	0	1	1	1	?	0	-	-	
10	0	1	0	0	1	?	0	-	-	
13	0	1	1	0	0	?	0	-	-	
14	0	1	1	0	1	?	0	-	-	
15	0	1	1	1	0	?	0	-	-	
16	0	1	1	1	1	?	0	-	-	
18	1	0	0	0	1	?	0	-	-	
19	1	0	0	1	0	?	0	-	-	
20	1	0	0	1	1	?	0	-	-	
22	1	0	1	0	1	?	0	-	-	
23	1	0	1	1	0	?	0	-	-	
24	1	0	1	1	1	?	0	-	-	
27	1	1	0	1	0	?	0	-	-	
29	1	1	1	0	0	?	0	-	-	
30	1	1	1	0	1	?	0	-	-	
31	1	1	1	1	0	?	0	-	-	
32	1	1	1	1	1	?	0	-	-	

### 3.1.2 Constructing the truth table for the negated outcome

```
> ttDS2019neg <- truthTable(DS, outcome = "~OUT", conditions = "CON,
LIBDEM, THAW, NEIGH, CATH", incl.cut = 0.7, complete = TRUE, show.cases
= TRUE, sort.by = "out, n")
> ttDS2019neg
```

```
OUT: output value
n: number of cases in configuration
incl: sufficiency inclusion score
PRI: proportional reduction in inconsistency
```

	CON	LIBDEM	THAW	NEIGH	CATH	OUT	n	incl	PRI	cases
1	0	0	0	0	0	1	6	1.000	1.000	1,4,10,11,12,20
2	0	0	0	0	1	1	2	1.000	1.000	18,21
12	0	1	0	1	1	1	2	1.000	1.000	2,13
11	0	1	0	1	0	1	1	1.000	1.000	23
17	1	0	0	0	0	1	1	1.000	1.000	19
26	1	1	0	0	1	1	1	1.000	1.000	14
28	1	1	0	1	1	1	1	1.000	1.000	3
9	0	1	0	0	0	0	4	0.000	0.000	6,8,16,22
25	1	1	0	0	0	0	4	0.000	0.000	7,9,15,24
5	0	0	1	0	0	0	2	0.000	0.000	5,25
21	1	0	1	0	0	0	1	0.000	0.000	17
3	0	0	0	1	0	?	0	-	-	
4	0	0	0	1	1	?	0	-	-	
6	0	0	1	0	1	?	0	-	-	
7	0	0	1	1	0	?	0	-	-	
8	0	0	1	1	1	?	0	-	-	
10	0	1	0	0	1	?	0	-	-	
13	0	1	1	0	0	?	0	-	-	
14	0	1	1	0	1	?	0	-	-	
15	0	1	1	1	0	?	0	-	-	
16	0	1	1	1	1	?	0	-	-	
18	1	0	0	0	1	?	0	-	-	
19	1	0	0	1	0	?	0	-	-	
20	1	0	0	1	1	?	0	-	-	
22	1	0	1	0	1	?	0	-	-	
23	1	0	1	1	0	?	0	-	-	
24	1	0	1	1	1	?	0	-	-	
27	1	1	0	1	0	?	0	-	-	
29	1	1	1	0	0	?	0	-	-	
30	1	1	1	0	1	?	0	-	-	
31	1	1	1	1	0	?	0	-	-	
32	1	1	1	1	1	?	0	-	-	

## 3.2 Minimization

### 3.2.1 Complex solution, OUT=1

```
> consSOL <- minimize(ttDS2019, details=TRUE)
> consSOL
```

```
n OUT = 1/0/C: 11/14/0
  Total      : 25
```

Number of multiple-covered cases: 0

M1: LIBDEM\*thaw\*neigh\*cath + libdem\*THAW\*neigh\*cath <=> OUT

	incls	PRI	covS	covU	cases
1 LIBDEM*thaw*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
2 libdem*THAW*neigh*cath	1.000	1.000	0.273	0.273	5,25; 17
M1	1.000	1.000	1.000		

```
> consSOL$PIchart
```

	5	9	21	25
LIBDEM*thaw*neigh*cath	-	x	-	x
libdem*THAW*neigh*cath	x	-	x	-

### 3.2.2 Most parsimonious solution, OUT=1

```
> mps <- minimize(ttDS2019, include = "?", details = TRUE)
> mps
```

```
n OUT = 1/0/C: 11/14/0
  Total      : 25
```

Number of multiple-covered cases: 0

M1: THAW + LIBDEM\*neigh\*cath <=> OUT

	incls	PRI	covS	covU	cases
1 THAW	1.000	1.000	0.273	0.273	5,25; 17
2 LIBDEM*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
M1	1.000	1.000	1.000		

#### 3.2.2.1 Simplifying assumptions, mps, OUT=1

```
> mps$SA
$M1
```

	CON	LIBDEM	THAW	NEIGH	CATH
6	0	0	1	0	1
7	0	0	1	1	0
8	0	0	1	1	1

```

13  0    1    1    0    0
14  0    1    1    0    1
15  0    1    1    1    0
16  0    1    1    1    1
22  1    0    1    0    1
23  1    0    1    1    0
24  1    0    1    1    1
29  1    1    1    0    0
30  1    1    1    0    1
31  1    1    1    1    0
32  1    1    1    1    1

```

### 3.2.3 Enhanced parsimonious solution, OUT=1

```

> eps <- minimize(ttDS2019, include = "?", exclude =
c(3,4,13,14,15,16,18,19,20,29,30,31,32), details = TRUE)
> eps

```

```

n OUT = 1/0/C: 11/14/0
  Total      : 25

```

Number of multiple-covered cases: 0

M1: libdem\*THAW + LIBDEM\*thaw\*neigh\*cath <=> OUT

	incls	PRI	covS	covU	cases
1 libdem*THAW	1.000	1.000	0.273	0.273	5,25; 17
2 LIBDEM*thaw*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
M1	1.000	1.000	1.000		

#### 3.2.3.1 Simplifying assumptions, eps, OUT=1

```

> eps$SA
$M1
  CON LIBDEM THAW NEIGH CATH
6   0     0    1    0    1
7   0     0    1    1    0
8   0     0    1    1    1
22  1     0    1    0    1
23  1     0    1    1    0
24  1     0    1    1    1

```

### 3.2.4 Enhanced intermediate solution, OUT=1

```

> eSOL <- minimize(ttDS2019, include = "?", exclude =
c(3,4,13,14,15,16,18,19,20,29,30,31,32), dir.exp = "-,1,-,0,0", details =
TRUE)
> eSOL

```

```

n OUT = 1/0/C: 11/14/0
  Total      : 25

```

From C1P1:



Number of multiple-covered cases: 0

M1: LIBDEM\*thaw\*neigh\*cath + libdem\*THAW\*neigh\*cath <=> OUT

		inclS	PRI	covS	covU	cases
1	LIBDEM*thaw*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
2	libdem*THAW*neigh*cath	1.000	1.000	0.273	0.273	5,25; 17
-----						
	M1	1.000	1.000	1.000		

### 3.2.4.1 Simplifying assumptions, easy and difficult counterfactuals, eis, OUT=1

*Note: all simplifying assumptions are difficult counterfactuals. Therefore, no minimization was possible, and the enhanced intermediary solution is identical with the complex solution.*

```
> eSOL$SA
```

```
$M1
```

```
      CON LIBDEM THAW NEIGH CATH
6      0      0      1      0      1
7      0      0      1      1      0
8      0      0      1      1      1
22     1      0      1      0      1
23     1      0      1      1      0
24     1      0      1      1      1
```

```
> eSOL$i.sol$C1P1$EC
```

```
[1] CON      LIBDEM THAW      NEIGH  CATH
<0 Zeilen> (oder row.names mit Länge 0)
```

```
> eSOL$i.sol$C1P1$DC
```

```
      CON LIBDEM THAW NEIGH CATH
6      0      0      1      0      1
7      0      0      1      1      0
8      0      0      1      1      1
22     1      0      1      0      1
23     1      0      1      1      0
24     1      0      1      1      1
```

### 3.2.5 Complex solution, OUT=0

```
> consSOLneg <- minimize(ttDS2019neg, details=TRUE)
> consSOLneg
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

Number of multiple-covered cases: 6

```
M1: con*LIBDEM*thaw*NEIGH + con*libdem*thaw*neigh + CON*LIBDEM*thaw*CATH +
    libdem*thaw*neigh*cath <=> out
```

	incls	PRI	covS	covU	cases
1 con*LIBDEM*thaw*NEIGH	1.000	1.000	0.214	0.214	23; 2,13
2 con*libdem*thaw*neigh	1.000	1.000	0.571	0.143	1,4,10,11,12,20; 18,21
3 CON*LIBDEM*thaw*CATH	1.000	1.000	0.143	0.143	14; 3
4 libdem*thaw*neigh*cath	1.000	1.000	0.500	0.071	1,4,10,11,12,20; 19
M1	1.000	1.000	1.000		

```
> consSOLneg$PIchart
```

	1	2	11	12	17	26	28
con*LIBDEM*thaw*NEIGH	-	-	x	x	-	-	-
con*libdem*thaw*neigh	x	x	-	-	-	-	-
CON*LIBDEM*thaw*CATH	-	-	-	-	-	x	x
libdem*thaw*neigh*cath	x	-	-	-	x	-	-
LIBDEM*thaw*NEIGH*CATH	-	-	-	x	-	-	x

### 3.2.6 Most parsimonious solution, OUT=0

```
> mpsneg <- minimize(ttDS2019neg, include = "?", details = TRUE)
> mpsneg
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

Number of multiple-covered cases: 5

```
M1: NEIGH + CATH + libdem*thaw <=> out
```

	incls	PRI	covS	covU	cases
1 NEIGH	1.000	1.000	0.286	0.071	23; 2,13; 3
2 CATH	1.000	1.000	0.429	0.071	18,21; 2,13; 14; 3
3 libdem*thaw	1.000	1.000	0.643	0.500	1,4,10,11,12,20; 18,21; 19
M1	1.000	1.000	1.000		

### 3.2.6.1 Simplifying assumptions, mps, OUT=0

```
> mpsneg$SA
$M1
  CON LIBDEM THAW NEIGH CATH
3    0      0    0     1    0
4    0      0    0     1    1
6    0      0    1     0    1
7    0      0    1     1    0
8    0      0    1     1    1
10   0      1    0     0    1
14   0      1    1     0    1
15   0      1    1     1    0
16   0      1    1     1    1
18   1      0    0     0    1
19   1      0    0     1    0
20   1      0    0     1    1
22   1      0    1     0    1
23   1      0    1     1    0
24   1      0    1     1    1
27   1      1    0     1    0
30   1      1    1     0    1
31   1      1    1     1    0
32   1      1    1     1    1
```

### 3.2.7 Enhanced parsimonious solution, OUT=0

```
> epsneg <- minimize(ttDS2019neg, include = "?", exclude =
c(13,14,15,16,29,30,31,32), details = TRUE)
> epsneg
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

Number of multiple-covered cases: 5

M1: libdem\*thaw + thaw\*NEIGH + thaw\*CATH <=> out

		inclS	PRI	covS	covU	cases
1	libdem*thaw	1.000	1.000	0.643	0.500	1,4,10,11,12,20; 18,21; 19
2	thaw*NEIGH	1.000	1.000	0.286	0.071	23; 2,13; 3
3	thaw*CATH	1.000	1.000	0.429	0.071	18,21; 2,13; 14; 3
-----						
	M1	1.000	1.000	1.000		

#### 3.2.7.1 Simplifying assumptions, eps, OUT=0

```
> epsneg$SA
$M1
  CON LIBDEM THAW NEIGH CATH
3    0      0    0     1    0
4    0      0    0     1    1
10   0      1    0     0    1
18   1      0    0     0    1
19   1      0    0     1    0
20   1      0    0     1    1
27   1      1    0     1    0
```

### 3.2.8 Theory-guided enhanced intermediate solution, OUT=0

```
> teSOLneg <- minimize(ttDS2019neg, include = "?", dir.exp = "-,0,0,-,-",
exclude = c(13,14,15,16,23,27,29,30,31,32), details = TRUE)
> teSOLneg
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

From C1P1:

Number of multiple-covered cases: 0

M1: con\*thaw\*NEIGH + CON\*thaw\*CATH + libdem\*thaw\*neigh <=> out

	inclS	PRI	covS	covU	cases
1 con*thaw*NEIGH	1.000	1.000	0.214	0.214	23; 2,13
2 CON*thaw*CATH	1.000	1.000	0.143	0.143	14; 3
3 libdem*thaw*neigh	1.000	1.000	0.643	0.643	1,4,10,11,12,20; 18,21; 19
M1	1.000	1.000	1.000		

#### 3.2.8.1 Simplifying assumptions, easy and difficult counterfactuals, teis, OUT=0

```
> teSOLneg$SA
$M1
```

	CON	LIBDEM	THAW	NEIGH	CATH
3	0	0	0	1	0
4	0	0	0	1	1
10	0	1	0	0	1
18	1	0	0	0	1
19	1	0	0	1	0
20	1	0	0	1	1

```
> teSOLneg$i.sol$C1P1$EC
```

	CON	LIBDEM	THAW	NEIGH	CATH
3	0	0	0	1	0
4	0	0	0	1	1
18	1	0	0	0	1
20	1	0	0	1	1

```
> teSOLneg$i.sol$C1P1$DC
```

	CON	LIBDEM	THAW	NEIGH	CATH
10	0	1	0	0	1
19	1	0	0	1	0

#### 3.2.8.2 Robustness of conjunctural expectations, teis, OUT=0

To assess the robustness of the conjunctural directional expectations, I compare the enhanced intermediate solution (without conjunctural expectations) with the theory-guided enhanced intermediate solution.

```
> teSOLnegA <- minimize(ttDS2019neg, include = "?", dir.exp = "-,0,0,-,-",
exclude = c(13,14,15,16,29,30,31,32), details = TRUE)
> teSOLnegA
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

From C1P1:

Number of multiple-covered cases: 0

```
M1:    con*thaw*NEIGH + CON*thaw*CATH + libdem*thaw*neigh <=> out
```

	incls	PRI	covS	covU	cases	
1	con*thaw*NEIGH	1.000	1.000	0.214	0.214	23; 2,13
2	CON*thaw*CATH	1.000	1.000	0.143	0.143	14; 3
3	libdem*thaw*neigh	1.000	1.000	0.643	0.643	1,4,10,11,12,20; 18,21; 19
M1		1.000	1.000	1.000		

```
> teSOLnegA$SA
$M1
```

	CON	LIBDEM	THAW	NEIGH	CATH
3	0	0	0	1	0
4	0	0	0	1	1
10	0	1	0	0	1
18	1	0	0	0	1
19	1	0	0	1	0
20	1	0	0	1	1
27	1	1	0	1	0

```
> teSOLnegA$i.sol$C1P1$EC
```

	CON	LIBDEM	THAW	NEIGH	CATH
3	0	0	0	1	0
4	0	0	0	1	1
18	1	0	0	0	1
20	1	0	0	1	1

```
> teSOLnegA$i.sol$C1P1$DC
```

	CON	LIBDEM	THAW	NEIGH	CATH
10	0	1	0	0	1
19	1	0	0	1	0
27	1	1	0	1	0

*Note: The solutions with or without conjunctural expectations are the same. One of the remainders (27, corresponds to 28 in the manuscript) is a simplifying assumption, but a difficult one, thus not entering the minimization.*

### 3.3 Testing for model ambiguity

Model ambiguity arises when different solutions are possible with the data at hand. Because I rely mostly on the (theory-guided) enhanced intermediary solution in my interpretations, I checked for model ambiguity for these solutions.

#### 3.3.1 PI Charts

PI Charts identify non-essential prime implicants. A prime implicant is redundant if all primitive expressions (the columns) are covered even in its absence. Therefore, redundant prime implicants can be removed to arrive at a more parsimonious solution. A possible problem arises if alternative ways of removing prime implicants are possible.

The PI chart for the enhanced intermediate solution of the positive outcome shows that two alternative models are possible (lines 1 and 4; or lines 1,2,3). In the case of the complex solution, that is otherwise homonymous with the enhanced intermediate solution, no ambiguity arises. The reasons for choosing the model 1,4 and the implications this has for the analysis are explained in the manuscript.

The PI chart for the enhanced intermediate solution of the positive outcome is:

```
> eSOL$PIchart
```

	5	9	21	25
libdem*THAW	x	-	x	-
con*LIBDEM*thaw*neigh	-	x	-	-
CON*LIBDEM*thaw*cath	-	-	-	x
LIBDEM*thaw*neigh*cath	-	x	-	x

#### PI chart for the complex solution

```
> consOL$PIchart
```

	5	9	21	25
LIBDEM*thaw*neigh*cath	-	x	-	x
libdem*THAW*neigh*cath	x	-	x	-

PI chart for the theory-guided enhanced intermediate solution of the negated outcome:

```
> teSOLneg$PIchart
```

	1	2	11	12	17	26	28
libdem*thaw	x	x	-	-	x	-	-
libdem*CATH	-	x	-	-	-	-	-
thaw*CATH	-	x	-	x	-	x	x
con*thaw*NEIGH	-	-	x	x	-	-	-

*Here, only the one prime implicant (libdem\*CATH) is redundant, leaving no ambiguity.*

### 3.3.2 Row dominance

The second test tries the effects of using and non-using the option “row dominance” in the minimization procedure. Row dominance eliminates redundant PI charts before minimization. Therefore, I compare the solutions when row dominance is enabled with the solutions when it is disabled. (The default setting in the QCA package is `row.dom = TRUE`). As the output shows, the two solutions are identical.

```
> eSOLROWtrue <- minimize(ttDS2019, include = "?", exclude =  
c(3,4,13,14,15,16,18,19,20,29,30,31,32), dir.exp = "-,1,-,0,0", row.dom =  
TRUE, details = TRUE)
```

```
> eSOLROWtrue
```

```
n OUT = 1/0/C: 11/14/0  
Total      : 25
```

```
From C1P1:
```

```
Number of multiple-covered cases: 0
```

```
M1: LIBDEM*thaw*neigh*cath + libdem*THAW*neigh*cath <=> OUT
```

	incls	PRI	covS	covU	cases
1 LIBDEM*thaw*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
2 libdem*THAW*neigh*cath	1.000	1.000	0.273	0.273	5,25; 17
M1	1.000	1.000	1.000		

```
> eSOLROWFALSE <- minimize(ttDS2019, include = "?", exclude =  
c(3,4,13,14,15,16,18,19,20,29,30,31,32), dir.exp = "-,1,-,0,0", row.dom =  
FALSE, details = TRUE)
```

```
> eSOLROWFALSE
```

```
n OUT = 1/0/C: 11/14/0  
Total      : 25
```

```
From C1P1:
```

```
Number of multiple-covered cases: 0
```

```
M1: LIBDEM*thaw*neigh*cath + libdem*THAW*neigh*cath <=> OUT
```

	incls	PRI	covS	covU	cases
1 LIBDEM*thaw*neigh*cath	1.000	1.000	0.727	0.727	6,8,16,22; 7,9,15,24
2 libdem*THAW*neigh*cath	1.000	1.000	0.273	0.273	5,25; 17
M1	1.000	1.000	1.000		

```
> teSOLnegROWtrue <- minimize(ttDS2019neg, include = "?", dir.exp = "-
,0,0,-,-", exclude = c(13,14,15,16,23,27,29,30,31,32), row.dom = TRUE,
details = TRUE)
> teSOLnegROWtrue
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

From C1P1:

Number of multiple-covered cases: 0

```
M1:    con*thaw*NEIGH + CON*thaw*CATH + libdem*thaw*neigh <=> out
```

	incls	PRI	covS	covU	cases	
1	con*thaw*NEIGH	1.000	1.000	0.214	0.214	23; 2,13
2	CON*thaw*CATH	1.000	1.000	0.143	0.143	14; 3
3	libdem*thaw*neigh	1.000	1.000	0.643	0.643	1,4,10,11,12,20; 18,21; 19
M1		1.000	1.000	1.000		

```
> teSOLnegROWFALSE <- minimize(ttDS2019neg, include = "?", dir.exp = "-
,0,0,-,-", exclude = c(13,14,15,16,23,27,29,30,31,32), row.dom = FALSE,
details = TRUE)
```

```
> teSOLnegROWFALSE
```

```
n OUT = 1/0/C: 14/11/0
  Total      : 25
```

From C1P1:

Number of multiple-covered cases: 0

```
M1:    con*thaw*NEIGH + CON*thaw*CATH + libdem*thaw*neigh <=> out
```

	incls	PRI	covS	covU	cases	
1	con*thaw*NEIGH	1.000	1.000	0.214	0.214	23; 2,13
2	CON*thaw*CATH	1.000	1.000	0.143	0.143	14; 3
3	libdem*thaw*neigh	1.000	1.000	0.643	0.643	1,4,10,11,12,20; 18,21; 19
M1		1.000	1.000	1.000		