Logic Synthesis & Verification, Fall 2012 National Taiwan University

Problem Set 2

Due on 2012/10/31 before lecture

1 [Cofactor and QBF]

- (a) (6%) Given two arbitrary Boolean functions f and g and a Boolean variable v, prove that $(\neg f)_v = \neg(f_v)$ and $(f \langle op \rangle g)_v = (f_v) \langle op \rangle (g_v)$ for $\langle op \rangle = \{\lor, \oplus\}$.
- (b) (12%) Prove or disprove the following implications:

$$\forall x, \exists y. f(x, y, z) \leftrightarrow \exists y, \forall x. f(x, y, z) \tag{1}$$

 $\forall x.(f(x,y) \land g(x,y)) \leftrightarrow (\forall x.f(x,y)) \land (\forall x.g(x,y))$ (2)

$$\forall x.(f(x,y) \lor g(x,y)) \leftrightarrow (\forall x.f(x,y)) \lor (\forall x.g(x,y))$$
(3)

(c) (12%) For an arbitrary Boolean function $f(x_1, \ldots, x_n)$, let a Boolean function $g(x_1, \ldots, x_{n-1})$ satisfy

 $\forall x_n f(x_1, \dots, x_n) = f(x_1, \dots, x_{n-1}, g(x_1, \dots, x_{n-1})).$

Express the on-set, off-set, and don't-care-set of g in terms of function f.

2 [BDD Operation]

Let $f = \neg ab \neg c \lor a \neg cd \lor ac \neg d$ and $g = c \oplus d \oplus e$.

- (a) (10%) Draw the (shared) ROBDDs of f and g under variable ordering a < b < c < d < e (with a on top).
- (b) (10%) Compute the ROBDD of COMPOSE(f, c, g).

3 [SAT Solving]

In pseudo Boolean constraint solving, one method is to translate the constraints into a CNF formula for SAT solving. Consider the linear inequality $5x_1 + 3x_2 + x_3 + x_4 + x_5 \ge 6$, where x_i 's are Boolean variables and "+" is arithmetic addition.

- (a) (10%) Build an ROBDD (variable ordering $x_1 < x_2 < x_3 < x_4 < x_5$) that characterizes the set of feasible solutions to the inequality.
- (b) (10%) Treat the above ROBDD as a network of multiplexors and translate it to a CNF formula.
- (c) (10%) Show that there exists a linear inequality whose solution-characterizing ROBDD has nodes exponential in the number of variables.

4 [SAT Solving]

Consider SAT solving the CNF formula consisting of the following 8 clauses

$$C_1 = (a + b + c), C_2 = (a + b' + d), C_3 = (a + b + c' + d'),$$

$$C_4 = (b' + c' + d), C_5 = (a + b + d), C_6 = (a' + b' + c),$$

$$C_7 = (a' + b + d), C_8 = (a + c + d'), C_9 = (a + b' + d').$$

- (a) (10%) Apply implication and conflict-based learning in solving the above CNF formula. Assume the decision order follows a, b, c, and then d; assume each variable is assigned 0 first and then 1. Whenever a conflict occurs, draw the implication graph and enumerate all possible learned clauses under the Unique Implication Point (UIP) principle. (In your implication graphs, annotate each vertex with "variable = value@decision_level", e.g., "b = 0@2", and annotate each edge with the clause that implication happens.) If there are multiple UIP learned clauses for a conflict, use the one with the UIP closest to the conflict in the implication graph.
- (b) (10%) The **resolution** between two clauses $C_i = (C_i^* + x)$ and $C_j = (C_j^* + x')$ (where C_i^* and C_j^* are sub-clauses of C_i and C_j , respectively) is the process of generating their **resolvent** $(C_1^* + C_j^*)$. The resolution is often denoted as

$$\frac{(C_i^* + x) \quad (C_j^* + x')}{(C_1^* + C_j^*)}$$

A fact is that a learned clause in SAT solving can be derived by a few resolution steps. Show how that the learned clauses of (a) can be obtained by resolution with respect to their implication graphs.