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import random
import csv
def g_0(n):
    return ("?",)*n

def s_0(n):
    return (' $\phi$ ',)*n
def more_general(h1, h2):
    more_general_parts = []
    for x, y in zip(h1, h2):
        mg = x == "?" or (x != " $\phi$ " and (x == y or y == " $\phi$ "))
        more_general_parts.append(mg)
    return all(more_general_parts)
def fulfills(example, hypothesis):
    ### the implementation is the same as for hypotheses:
    return more_general(hypothesis, example)

def min_generalizations(h, x):
    h_new = list(h)
    for i in range(len(h)):
        if not fulfills(x[i:i+1], h[i:i+1]):
            h_new[i] = '?' if h[i] != ' $\phi$ ' else x[i]
    return [tuple(h_new)]
def min_specializations(h, domains, x):
    results = []
    for i in range(len(h)):
        if h[i] == "?":
            for val in domains[i]:
                if x[i] != val:
                    h_new = h[:i] + (val,) + h[i+1:]
                    results.append(h_new)
        elif h[i] != " $\phi$ ":
            h_new = h[:i] + (' $\phi$ ',) + h[i+1:]
            results.append(h_new)
    return results
with open('trainingexamples.csv') as csvFile:
    examples = [tuple(line) for line in csv.reader(csvFile)]
def get_domains(examples):
    d = [set() for i in examples[0]]
    for x in examples:
        for i, xi in enumerate(x):
            d[i].add(xi)
    return [list(sorted(x)) for x in d]
get_domains(examples)

def candidate_elimination(examples):
    domains = get_domains(examples)[-1]

    G = set([g_0(len(domains))])
    S = set([s_0(len(domains))])
    i = 0
    print("\n G[{0}]:".format(i), G)
    print("\n S[{0}]:".format(i), S)
    for xcx in examples:
        i = i + 1
        x, cx = xcx[:-1], xcx[-1] # Splitting data into attributes and decisions

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    if cx == 'Y': # x is positive example
        G = {g for g in G if fulfills(x, g)}
        S = generalize_S(x, G, S)
    else: # x is negative example
        S = {s for s in S if not fulfills(x, s)}
        G = specialize_G(x, domains, G, S)
    print("\n G[{0}]:".format(i), G)
    print("\n S[{0}]:".format(i), S)
    return
def generalize_S(x, G, S):
    S_prev = list(S)
    for s in S_prev:
        if s not in S:
            continue
        if not fulfills(x, s):
            S.remove(s)
            Splus = min_generalizations(s, x)
            ## keep only generalizations that have a counterpart in G
            S.update([h for h in Splus if any([more_general(g, h)
                                             for g in G])])
            ## remove hypotheses less specific than any other in S
            S.difference_update([h for h in S if
                                any([more_general(h, h1)
                                     for h1 in S if h != h1])])
    return S
def specialize_G(x, domains, G, S):
    G_prev = list(G)
    for g in G_prev:
        if g not in G:
            continue
        if fulfills(x, g):
            G.remove(g)
            Gminus = min_specializations(g, domains, x)
            ## keep only specializations that have a counterpart in S
            G.update([h for h in Gminus if any([more_general(h, s)
                                             for s in S])])
            ## remove hypotheses less general than any other in G
            G.difference_update([h for h in G if
                                any([more_general(g1, h)
                                     for g1 in G if h != g1])])
    return G
candidate_elimination(examples)

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