# ILP formulation and LP relaxation

Integer Linear Programming based approach

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1 Set cover to ILP reduction



### Set cover

Input : 
$$(U, S_1, S_2, \cdots, S_m)$$

- 1 introduce a variable  $x_i$  for every set  $S_i$
- 2  $x_i = 1$  when  $S_i$  is selected, and  $x_i = 0$  otherwise.



# LP formulation

# derive the linear programming relaxation

Minimize  $\sum_{i=1}^{m} x_i$ Subject to

- $x_i \le 1 \ \forall i \in \{1, 2, \cdots, m\}$
- $x_i \ge 0 \ \forall i \in \{1, 2, \cdots, m\}$

# Weighted LP optimization

## changing the cost method

Minimize  $\sum_{i=1}^{m} w_i.x_i$ Subject to

- $x_i \le 1 \ \forall i \in \{1, 2, \cdots, m\}$
- $x_i \ge 0 \ \forall i \in \{1, 2, \cdots, m\}$

# Weighted LP optimization

### optimal fractional solution $x^*$

$$x_i^* \in [0,1]$$

we cannot round the  $x_i^*$  to the nearest integer, because if an element u belongs to 100 sets, it could be that  $x_i^* = \frac{1}{100}$  for each of those sets, and we would be rounding all those numbers to zero, leaving the element u not covered.

# Weighted LP optimization

### Simply rounding doesn't work here

If we knew that every element u belongs to at most k sets, then we could round the numbers  $\geq \frac{1}{k}$  to 1, and the numbers  $\leq \frac{1}{k}$  to zero.

### k - approximation

Unfortunately, k could be very large, even  $\frac{n}{2}$ 

# Normalized Probability

# cosider $x_i^*$ as a probability

we can think of the solution  $x^*$  as describing a probability distribution over ways of choosing some of the subsets  $S_1, \dots, S_m$ , in which we choose  $S_1$  with probability  $x_1^*$ ,  $S_2$  with probability  $x_2^*$ , and so on.

# Randomization

### Algorithm RandomPick

- Input: values  $(x_1, \ldots, x_n)$  feasible for (3)
- $I := \emptyset$
- for i = 1 to n
  - with probability  $x_i$ , assign  $I := I \cup \{i\}$ , otherwise do nothing
- $\bullet$  return I

# RandomPick

#### Expected behaviour

Using this probabilistic process, the expected cost of the sets that we pick is  $i \sum_{i=1}^{m} w_i.x_i^*$ ,

#### Expected behaviour

Unfortunately, the solution that we construct could have a high probability of missing some of the elements

# Randomization

### Algorithm RandomizedRound

- 1. Input:  $x_1, \ldots, x_n$  feasible for (3)
- 2.  $I := \emptyset$
- 3. while there are elements u such that  $u \notin \bigcup_{i \in I} S_i$ 
  - for i := 1 to n
    - with probability  $x_i$ , assign  $I := I \cup \{i\}$ , otherwise do nothing
- 4. return I

# RandomidedRound

#### **Fact**

There is a probability at most  $e^{-100}$  that the while loop is executed for more than  $\ln |U| + 100$  times. In general, there is a probability at most  $e^{-k}$  that the while loop is executed form more than  $\ln |U| + k$  times.

## RandomidedRound

#### Fact

Fix any positive integer parameter t and any feasible solution  $(x_1,...,x_m)$  for LP formulation. Then the expected size of I in Algorithm RandomizedRound on input  $(x_1,...,x_m)$  after t iterations. (or at the end of the algorithm, if it ends in fewer than t iterations) is at most

$$t.\sum_{i=1}^m w_i.x_i$$

## RandomidedRound

#### **Fact**

Given an optimal solution  $(x_1^*, \dots, x_m^*)$  to (3), algorithm RandomizedRound outputs, with probability  $\geq$  .45, a feasible solution to the set cover problem that contains at most (2ln|U|+6) opt sets.