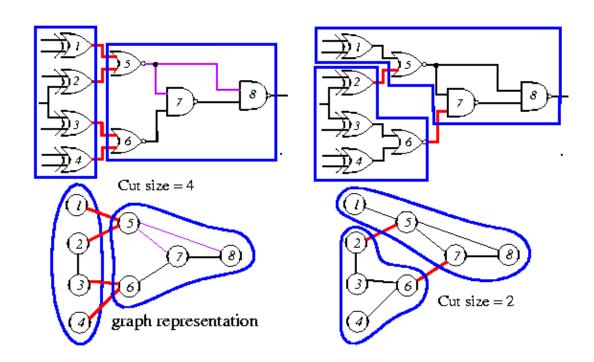
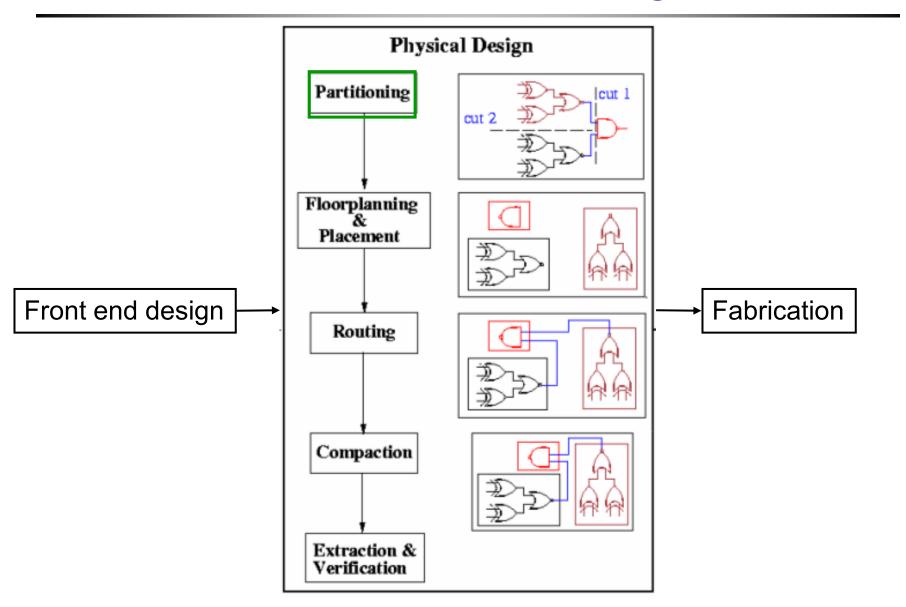
Unit 3: Circuit Partitioning

Course contents:

- Kernighan-Lin heuristic
- Simulated annealing based partitioning algorithm
- Fiduccia-Mattheyses heuristic

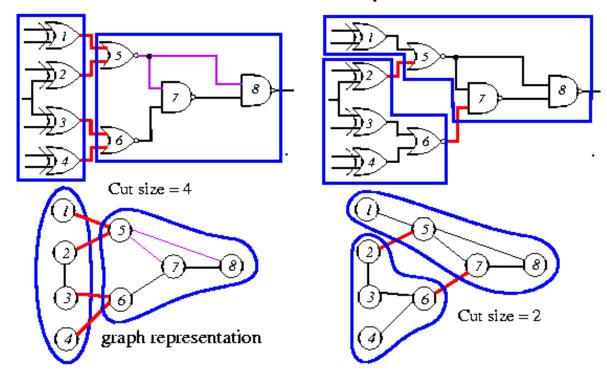


Position of Partitioning



Circuit Partitioning

- **Objective:** Partition a circuit into parts such that every component is within a prescribed range and the # of connections among the components is minimized.
 - More constraints are possible for some applications.
- Cutset? Cut size? Size of a component?



Problem Definition: Partitioning

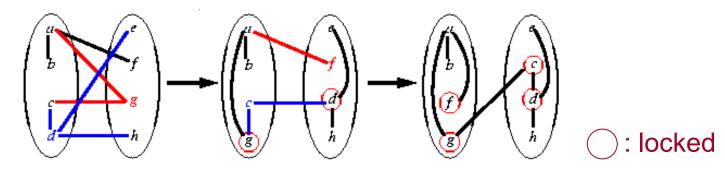
- k-way partitioning: Given a graph G(V, E), where each vertex $v \in V$ has a size s(v) and each edge $e \in E$ has a weight w(e), the problem is to divide the set V into k disjoint subsets $V_1, V_2, ..., V_k$, such that an objective function is optimized, subject to certain constraints.
- Bounded size constraint: The size of the *i*-th subset is bounded by B_i ($\sum_{v \in V_i} s(v) \leq B_i$).
 - Is the partition balanced?
- Min-cut cost between two subsets: Minimize $\sum_{\forall e=(u,v)\land p(u)\neq p(v)} w(e)$, where p(u) is the partition # of node u.
 - May not be balanced.
- The 2-way, balanced partitioning problem is NP-complete, even in its simple form with identical vertex sizes and unit edge weights.

Kernighan-Lin Algorithm

- Kernighan and Lin, "An efficient heuristic procedure for partitioning graphs," *The Bell System Technical Journal*, vol. 49, no. 2, Feb. 1970.
- An iterative, 2-way, balanced partitioning (bi-sectioning) heuristic.
- Till the cut size keeps decreasing
 - Vertex pairs which give the largest decrease or the smallest increase in cut size are exchanged.
 - These vertices are then **locked** (and thus are prohibited from participating in any further exchanges).
 - This process continues until all the vertices are locked.
 - Find the set with the largest partial sum for swapping.
 - Unlock all vertices.

K-L Algorithm: A Simple Example

Each edge has a unit weight.



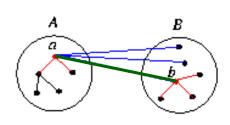
Step #	Vertex pair	Cost reduction	Cut cost
0	-	0	5
1	{d, g}	3	2
2	{c, f}	1	1
3	{b, h}	-2	3
4	{a, e}	-2	5

- Questions: How to compute cost reduction? What pairs to be swapped?
 - Consider the change of internal & external connections.

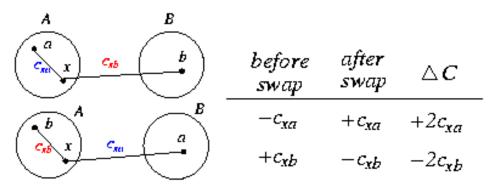
Properties

- Two sets A and B such that |A| = n = |B| and $A \cap B = \emptyset$.
- External cost of $a \in A$: $E_a = \sum_{v \in B} c_{av}$.
- Internal cost of $a \in A$: $I_a = \sum_{v \in A} c_{av}$.
- D-value of a vertex a: $D_a = E_a I_a$ (cost reduction for moving a).
- Cost reduction (gain) for swapping a and b: $g_{ab} = D_a + D_b 2c_{ab}$.
- If a ∈ A and b ∈ B are interchanged, then the new D-values, D', are given by

$$\begin{array}{rcl} D'_x & = & D_x + 2c_{xa} - 2c_{xb}, \forall x \in A - \{a\} \\ D'_y & = & D_y + 2c_{yb} - 2c_{ya}, \forall y \in B - \{b\}. \end{array}$$

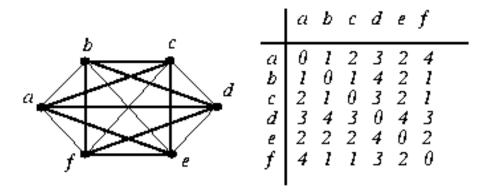


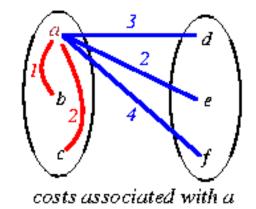
 $Gain_{a \rightarrow B}: D_a - c_{ab}$ $Gain_{b \rightarrow A}: D_b - c_{ab}$ Internal cost vs. External cost



updating D-values

K-L Algorithm: A Weighted Example





Initial cut cost = (3+2+4)+(4+2+1)+(3+2+1) = 22

• Iteration 1:

$$I_a = 1 + 2 = 3$$
; $E_a = 3 + 2 + 4 = 9$; $D_a = E_a - I_a = 9 - 3 = 6$
 $I_b = 1 + 1 = 2$; $E_b = 4 + 2 + 1 = 7$; $D_b = E_b - I_b = 7 - 2 = 5$
 $I_c = 2 + 1 = 3$; $E_c = 3 + 2 + 1 = 6$; $D_c = E_c - I_c = 6 - 3 = 3$
 $I_d = 4 + 3 = 7$; $E_d = 3 + 4 + 3 = 10$; $D_d = E_d - I_d = 10 - 7 = 3$
 $I_e = 4 + 2 = 6$; $E_e = 2 + 2 + 2 = 6$; $D_e = E_e - I_e = 6 - 6 = 0$
 $I_f = 3 + 2 = 5$; $E_f = 4 + 1 + 1 = 6$; $D_f = E_f - I_f = 6 - 5 = 1$

Computing the g Value

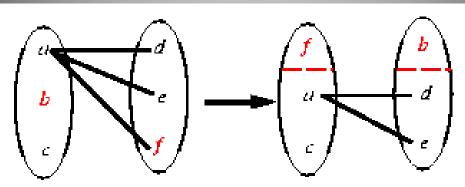
Iteration 1:

$$\begin{array}{lll} I_a=1+2=3; & E_a=3+2+4=9; & D_a=E_a-I_a=9-3=6\\ I_b=1+1=2; & E_b=4+2+1=7; & D_b=E_b-I_b=7-2=5\\ I_c=2+1=3; & E_c=3+2+1=6; & D_c=E_c-I_c=6-3=3\\ I_d=4+3=7; & E_d=3+4+3=10; & D_d=E_d-I_d=10-7=3\\ I_e=4+2=6; & E_e=2+2+2=6; & D_e=E_e-I_e=6-6=0\\ I_f=3+2=5; & E_f=4+1+1=6; & D_f=E_f-I_f=6-5=1 \end{array}$$

$$\begin{array}{lll} \bullet & g_{xy} = D_x + D_y - 2c_{xy}. \\ & g_{ad} & = & D_a + D_d - 2c_{ad} = 6 + 3 - 2 \times 3 = 3 \\ & g_{ae} & = & 6 + 0 - 2 \times 2 = 2 \\ & g_{af} & = & 6 + 1 - 2 \times 4 = -1 \\ & g_{bd} & = & 5 + 3 - 2 \times 4 = 0 \\ & g_{be} & = & 5 + 0 - 2 \times 2 = 1 \\ & g_{bf} & = & 5 + 1 - 2 \times 1 = 4 \; (maximum) \\ & g_{cd} & = & 3 + 3 - 2 \times 3 = 0 \\ & g_{ce} & = & 3 + 0 - 2 \times 2 = -1 \\ & g_{cf} & = & 3 + 1 - 2 \times 1 = 2 \\ \end{array}$$

• Swap b and f. $(\hat{g_1} = 4)$

Updating the D Value



• $D'_{x} = D_{x} + 2 c_{xp} - 2 c_{xq}$, $\forall x \in A - \{p\}$ (swap p and $q, p \in A, q \in B$)

$$D'_{a} = D_{a} + 2c_{ab} - 2c_{af} = 6 + 2 \times 1 - 2 \times 4 = 0$$

$$D'_{c} = D_{c} + 2c_{cb} - 2c_{cf} = 3 + 2 \times 1 - 2 \times 1 = 3$$

$$D'_{d} = D_{d} + 2c_{df} - 2c_{db} = 3 + 2 \times 3 - 2 \times 4 = 1$$

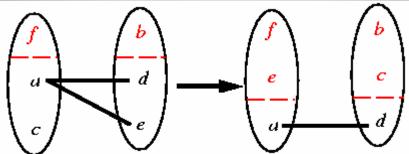
$$D'_{e} = D_{e} + 2c_{ef} - 2c_{eb} = 0 + 2 \times 2 - 2 \times 2 = 0$$

•
$$g_{xy} = D'_x + D'_y - 2c_{xy}$$

 $g_{ad} = D'_a + D'_d - 2c_{ad} = 0 + 1 - 2 \times 3 = -5$
 $g_{ae} = D'_a + D'_e - 2c_{ae} = 0 + 0 - 2 \times 2 = -4$
 $g_{cd} = D'_c + D'_d - 2c_{cd} = 3 + 1 - 2 \times 3 = -2$
 $g_{ce} = D'_c + D'_e - 2c_{ce} = 3 + 0 - 2 \times 2 = -1$ (maximum)

• Swap c and e. $(\hat{g}_2 = -1)$

Determining Swapping Pairs



•
$$D''_{x} = D'_{x} + 2c_{xp} - 2c_{xq}, \forall x \in A - \{p\}$$

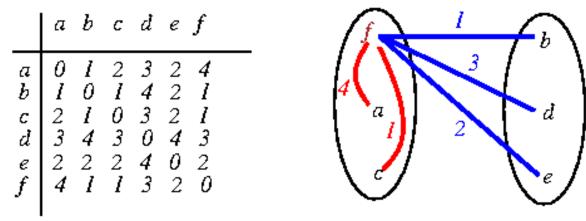
$$D_a'' = D_a' + 2c_{ac} - 2c_{ae} = 0 + 2 \times 2 - 2 \times 2 = 0$$

$$D_d'' = D_d' + 2c_{de} - 2c_{dc} = 1 + 2 \times 4 - 2 \times 3 = 3$$

•
$$g_{xy} = D_x'' + D_y'' - 2c_{xy}$$
.
 $g_{ad} = D_a'' + D_d'' - 2c_{ad} = 0 + 3 - 2 \times 3 = -3(\hat{g}_3 = -3)$

- Note that this step is redundant $(\sum_{i=1}^{n} \widehat{g_i} = 0)$.
- Summary: $\hat{g_1} = g_{bf} = 4$, $\hat{g_2} = g_{ce} = -1$, $\hat{g_3} = g_{ad} = -3$.
- Largest partial sum $\max \sum_{i=1}^{k} \widehat{g}_i = 4 \ (k = 1) \Rightarrow \text{Swap } b \text{ and } f$.

Next Iteration



Initial cut cost = (1+3+2)+(1+3+2)+(1+3+2) = 18(22-4)

- Iteration 2: Repeat what we did at Iteration 1 (Initial cost = 22-4 = 18).
- Summary: $\hat{g_1} = g_{ce} = -1$, $\hat{g_2} = g_{ab} = -3$, $\hat{g_3} = g_{fd} = 4$.
- Largest partial sum = $\max \sum_{i=1}^{k} \hat{g_i} = 0 \ (k = 3) \Rightarrow \text{Stop!}$

Kernighan-Lin Algorithm

```
Algorithm: Kernighan-Lin(G)
Input: G = (V, E), |V| = 2n.
Output: Balanced bi-partition A and B with "small" cut cost.
1 begin
2 Bipartition G into A and B such that |V_A| = |V_B|, |V_A| \cap V_B = \emptyset,
     and V_A \cup V_B = V.
3 repeat
4 Compute D_v, \forall v \in V.
5
    for i = 1 to n do
         Find a pair of unlocked vertices v_{ai} \in V_A and v_{bi} \in V_B whose
         exchange makes the largest decrease or smallest increase in
         cut cost:
         Mark v_{ai} and v_{bi} as locked, store the gain \hat{g_i}, and compute
7
         the new D_v, for all unlocked v \in V;
    Find k, such that G_k = \sum_{i=1}^k \widehat{g}_i is maximized;
8
     if G_k > 0 then
         Move v_{a1}, \ldots, v_{ak} from V_A to V_B and v_{b1}, \ldots, v_{bk} from V_B to V_A;
10
     Unlock v, \forall v \in V.
11
12 until G_k \leq 0;
13 end
```

Time Complexity

- Line 4: Initial computation of D: $O(n^2)$
- Line 5: The **for**-loop: *O*(*n*)
- The body of the loop: $O(n^2)$.
 - Lines 6—7: Step i takes $(n-i+1)^2$ time.
- Lines 4--11: Each pass of the repeat loop: $O(n^3)$.
- Suppose the repeat loop terminates after r passes.
- The total running time: $O(rn^3)$.
 - Polynomial-time algorithm?

Extensions of K-L Algorithm

- Unequal sized subsets (assume $n_1 < n_2$)
 - 1. Partition: $|A| = n_1$ and $|B| = n_2$.
 - 2. Add n_2 n_1 dummy vertices to set A. Dummy vertices have no connections to the original graph.
 - 3. Apply the Kernighan-Lin algorithm.
 - 4. Remove all dummy vertices.

Unequal sized "vertices"

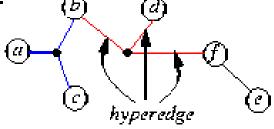
- 1. Assume that the smallest "vertex" has unit size.
- 2. Replace each vertex of size *s* with *s* vertices which are fully connected with edges of infinite weight.
- 3. Apply the Kernighan-Lin algorithm.

• k-way partition

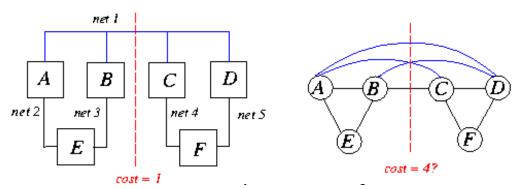
- Partition the graph into k equal-sized sets.
- 2. Apply the Kernighan-Lin algorithm for each pair of subsets.
- 3. Time complexity? Can be reduced by recursive bi-partition.
- How to handle hypergraphs?
 - Need to handle multi-terminal nets directly.

Coping with Hypergraph

• A hypergraph H = (N, L) consists of a set N of vertices and a set L of hyperedges, where each hyperedge corresponds to a **subset** N_i of distinct vertices with $|N_i| \ge 2$.



- Schweikert and Kernighan, "A proper model for the partitioning of electrical circuits," 9th Design Automation Workshop, 1972.
- For multi-terminal nets, **net cut** is a more accurate measurement for cut cost (i.e., deal with hyperedges).
 - {A, B, E}, {C, D, F} is a good partition.
 - Should not assign the same weight for all edges.

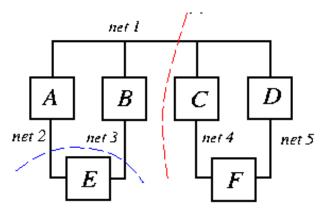


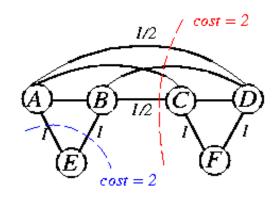
Net-Cut Model

• Let n(i) = # of cells associated with Net i.

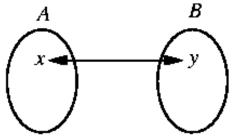
• Edge weight $w_{xy} = \frac{2}{n(i)}$ for an edge connecting cells x

and y.





• Easy modification of the K-L heuristic.



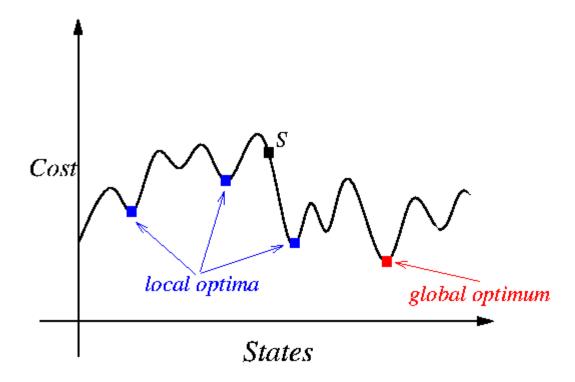
 D_x : gain in moving x to B

 D_{v} : gain in moving y to A

$$g_{xy} = D_x + D_y - Correction(x, y)$$

Simulated Annealing

- Kirkpatrick, Gelatt, and Vecchi, "Optimization by simulated annealing," Science, May 1983.
- Greene and Supowit, "Simulated annealing without rejected moves," ICCD-84.



Simulated Annealing Basics

- Non-zero probability for "up-hill" moves.
- Probability depends on
 - magnitude of the "up-hill" movement
 - total search time

$$Prob(S \rightarrow S') = \left\{ \begin{array}{ll} 1 & \text{if } \Delta C \leq 0 \quad / * "down-hill" \ moves * / \\ e^{-\Delta C} & \text{if } \Delta C > 0 \quad / * "up-hill" \ moves * / \end{array} \right.$$

- $\Delta C = cost(S') cost(S)$
- *T*: Control parameter (temperature)
- Annealing schedule: $T=T_0$, T_1 , T_2 , ..., where $T_i=r^i$ T_0 , r<1.

Generic Simulated Annealing Algorithm

```
1 begin
2 Get an initial solution S;
3 Get an initial temperature T > 0;
4 while not yet "frozen" do
    for 1 \le i \le P do
5
        Pick a random neighbor S' of S;
        \Delta \leftarrow cost(S') - cost(S);
       /* downhill move */
     if \Delta \leq 0 then S \leftarrow S'
        /* uphill move */
       if \Delta > 0 then S \leftarrow S' with probability e^{-T};
10 T \leftarrow rT; /* reduce temperature */
11 return S
12 end
```

Basic Ingredients for Simulated Annealing

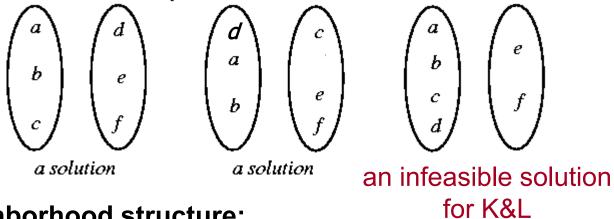
• Analogy:

Physical system	Optimization problem	
state	configuration	
energy	cost function	
ground state	optimal solution	
quenching	iterative improvement	
careful annealing	simulated annealing	

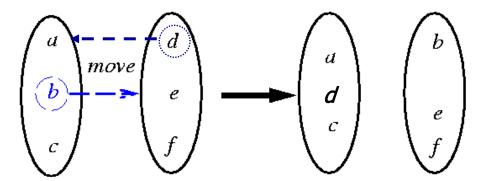
- Basic Ingredients for Simulated Annealing:
 - Solution space
 - Neighborhood structure
 - Cost function
 - Annealing schedule

Partition by Simulated Annealing

- Kirkpatrick, Gelatt, and Vecchi, "Optimization by simulated annealing," Science, May 1983.
- Solution space: set of all partitions



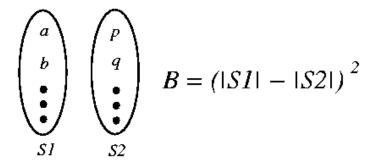
• Neighborhood structure:



Randomly swap a pair of cells from each side

Partition by Simulated Annealing (cont'd)

- Cost function: f = C
 - C: the partition cost as used before.



Annealing schedule:

- $T_n = r^n T_0, r = 0.9.$
- At each temperature, either
 - there are 10 accepted moves/cell on the average, or
 - ² # of attempts ≥ 100 x total # of cells.
- The system is "frozen" if very low acceptances at 3 consecutive temperatures.