

## Today we're going to cover

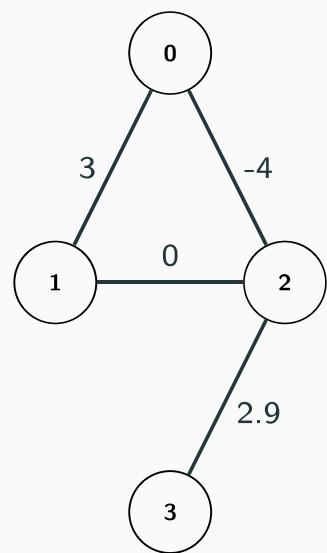
- Minimum spanning tree
- Shortest paths
- Some known graph problems
- Special graphs
  - Trees
  - Directed acyclic graphs
  - Bipartite graphs

## Weighted graphs

- Now the edges in our graphs may have weights, which could represent
  - the distance of the road represented by the edge
  - the cost of going over the edge
  - some capacity of the edge
- We can use a modified adjacency list to represent weighted graphs

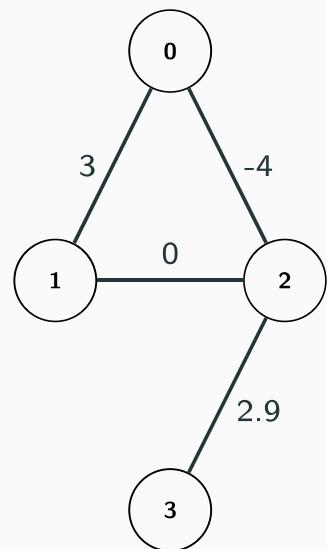
# Weighted graphs

```
struct edge {  
    int u, v;  
    int weight;  
  
    edge(int _u, int _v, int _w) {  
        u = _u;  
        v = _v;  
        weight = _w;  
    }  
};
```



## Weighted graphs

```
vector<edge> adj[4];  
  
adj[0].push_back(edge(0, 1, 3));  
adj[0].push_back(edge(0, 2, -4));  
  
adj[1].push_back(edge(1, 0, 3));  
adj[1].push_back(edge(1, 2, 0));  
  
adj[2].push_back(edge(2, 0, -4));  
adj[2].push_back(edge(2, 1, 0));  
adj[2].push_back(edge(2, 3, 2.9));  
  
adj[3].push_back(edge(3, 2, 2.9));
```



## Minimum spanning tree

- We have an undirected weighted graph
- The vertices along with a subset of the edges in the graph is called a spanning tree if
  - it forms a tree (i.e. does not contain a cycle) and
  - the tree spans all vertices (all vertices can reach all other vertices)
- The weight of a spanning tree is the sum of the weights of the edges in the subset
- We want to find a minimum spanning tree

## Minimum spanning tree

- Several greedy algorithms work
- Go through the edges in the graph in increasing order of weight
- Greedily pick an edge if it doesn't form a cycle (Union-Find can be used to keep track of when we would get a cycle)
- When we've gone through all edges, we have a minimum spanning tree
- This is Kruskal's algorithm
- Time complexity is  $O(E \log E)$

# Minimum spanning tree

```
bool edge_cmp(const edge &a, const edge &b) {
    return a.weight < b.weight;
}

vector<edge> mst(int n, vector<edge> edges) {
    union_find uf(n);
    sort(edges.begin(), edges.end(), edge_cmp);

    vector<edge> res;
    for (int i = 0; i < edges.size(); i++) {
        int u = edges[i].u,
            v = edges[i].v;

        if (uf.find(u) != uf.find(v)) {
            uf.unite(u, v);
            res.push_back(edges[i]);
        }
    }

    return res;
}
```

## Shortest paths

- We have a weighted graph (undirected or directed)
- Given two vertices  $u, v$ , what is the shortest path from  $u$  to  $v$ ?
- If all weights are the same, this can be solved with breadth-first search
- Of course, this is usually not the case...

## Shortest paths

- There are many known algorithms to find shortest paths
- Like breadth-first search, these algorithms usually find the shortest paths from a given start vertex to all other vertices
- Let's take a quick look at Dijkstra's algorithm, the Bellman-Ford algorithm and the Floyd-Warshall algorithm

## Dijkstra's algorithm

```
vector<edge> adj[100];
vector<int> dist(100, INF);

void dijkstra(int start) {
    dist[start] = 0;
    priority_queue<pair<int, int>,
        vector<pair<int, int> >,
        greater<pair<int, int> > > pq;
    pq.push(make_pair(dist[start], start));

    while (!pq.empty()) {
        int u = pq.top().second,
            d = pq.top().first;
        pq.pop();
        if (d > dist[u]) continue;
        for (int i = 0; i < adj[u].size(); i++) {
            int v = adj[u][i].v,
                w = adj[u][i].weight;
            if (w + dist[u] < dist[v]) {
                dist[v] = w + dist[u];
                pq.push(make_pair(dist[v], v));
            }
        }
    }
}
```

## Dijkstra's algorithm

- Time complexity is  $O(V \log E)$
- Note that this only works for non-negative weights

## Bellman-Ford algorithm

```
void bellman_ford(int n, int start) {  
  
    dist[start] = 0;  
  
    for (int i = 0; i < n - 1; i++) {  
        for (int u = 0; u < n; u++) {  
            for (int j = 0; j < adj[u].size(); j++) {  
                int v = adj[u][j].v;  
                int w = adj[u][j].weight;  
                dist[v] = min(dist[v], w + dist[u]);  
            }  
        }  
    }  
}
```

## Bellman-Ford algorithm

- Time complexity is  $O(V \times E)$
- Can be used to detect negative-weight cycles

## Floyd-Warshall algorithm

- What about using dynamic programming to compute shortest paths?
- Let  $\text{sp}(k, i, j)$  be the shortest path from  $i$  to  $j$  if we're only allowed to travel through the vertices  $0, \dots, k$
- Base case:  $\text{sp}(k, i, j) = 0$  if  $i = j$
- Base case:  $\text{sp}(-1, i, j) = \text{weight}[a][b]$  if  $(i, j) \in E$
- Base case:  $\text{sp}(-1, i, j) = \infty$
- $$\text{sp}(k, i, j) = \min \begin{cases} \text{sp}(k - 1, i, k) + \text{sp}(k - 1, k, j) \\ \text{sp}(k - 1, i, j) \end{cases}$$

## Floyd-Warshall algorithm

```
int dist[1000][1000];
int weight[1000][1000];

void floyd_marshall(int n) {
    for (int i = 0; i < n; i++) {
        for (int j = 0; j < n; j++) {
            dist[i][j] = i == j ? 0 : weight[i][j];
        }
    }

    for (int k = 0; k < n; k++) {
        for (int i = 0; i < n; i++) {
            for (int j = 0; j < n; j++) {
                dist[i][j] = min(dist[i][j], dist[i][k] + dist[k][j]);
            }
        }
    }
}
```

## Floyd-Warshall algorithm

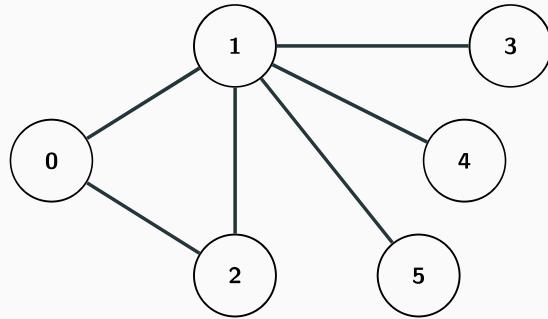
- Computes all-pairs shortest paths
- Time complexity is clearly  $O(n^3)$
- Very simple to code

## Known graph problems

- The problems we're dealing with very often ask us to solve some well known graph problem
- But usually it's well hidden in the problem statement
- Let's take a look at a few examples

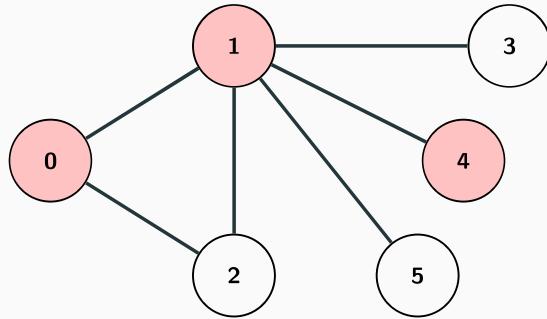
## Minimum vertex cover

- We have an unweighted undirected graph
- A vertex cover is a subset of the vertices  $S$ , such that for each edge  $(u, v)$  in the graph, either  $u$  or  $v$  (or both) are in  $S$



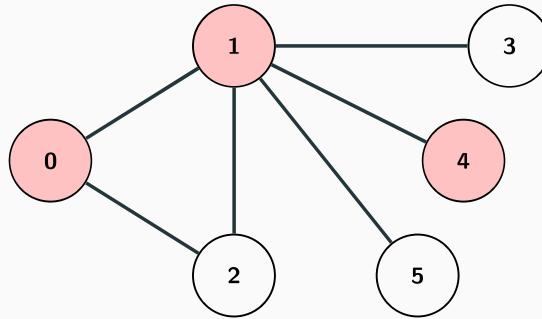
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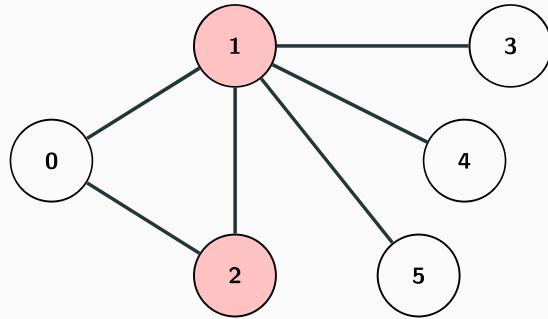
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- We want to find a vertex cover of minimum size

## Minimum vertex cover

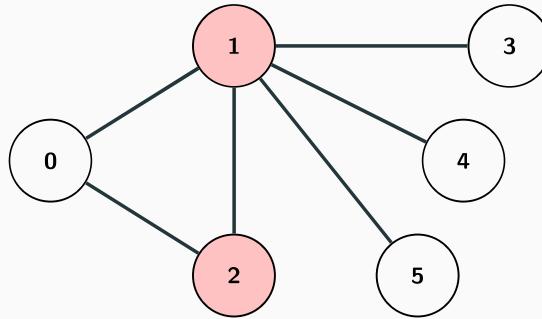
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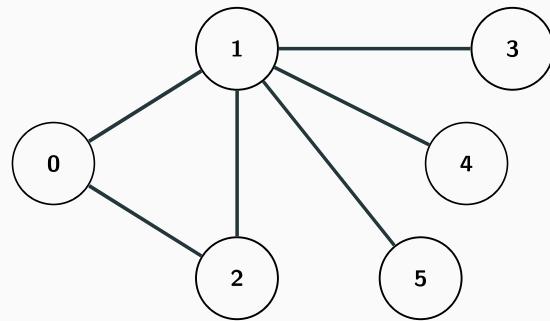
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- We want to find a vertex cover of minimum size
- NP-hard problem in general graphs

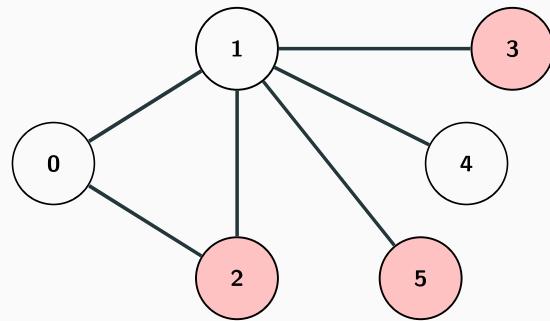
## Maximum independent set

- We have an unweighted undirected graph
- An independent set is a subset of the vertices  $S$ , such that no two vertices  $u, v$  in  $S$  are adjacent in the graph



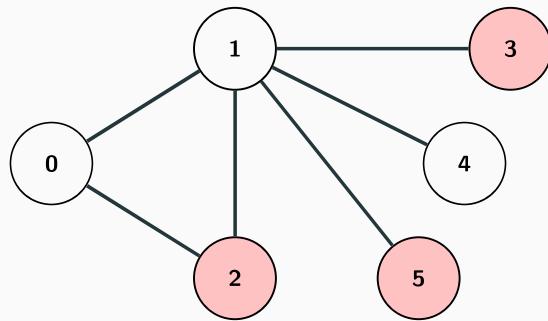
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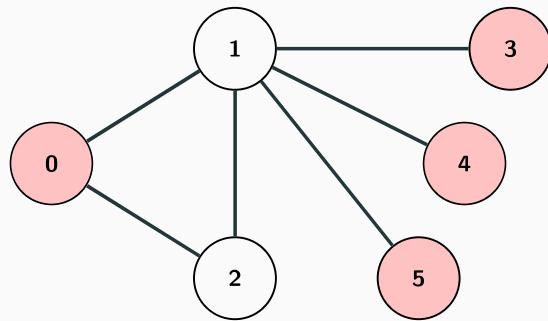
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- We want to find an independent set of maximum size

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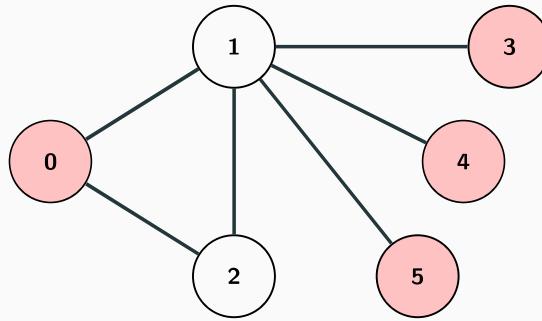
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## Maximum independent set

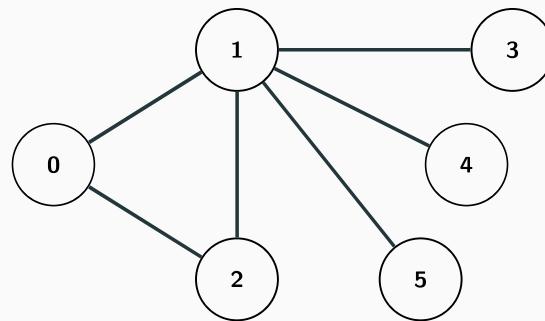
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- We want to find an independent set of maximum size
- NP-hard problem in general graphs

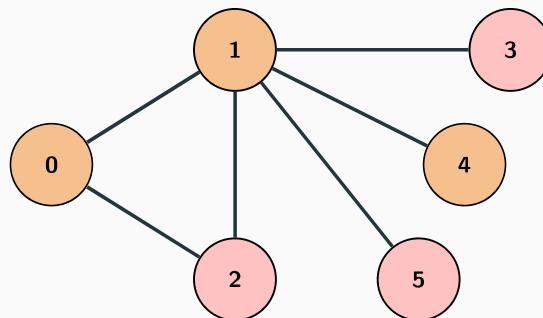
## Relation between MVC and MIS

- The previous two problems are very related
- A subset of the vertices is a vertex cover if and only if the complement of the set is an independent set



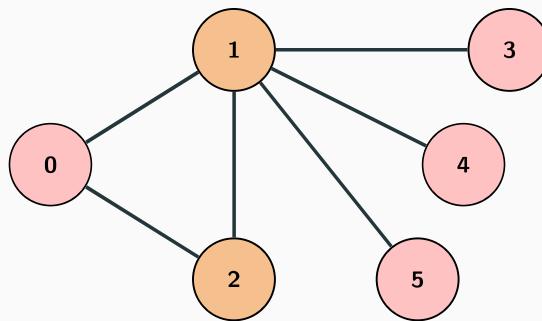
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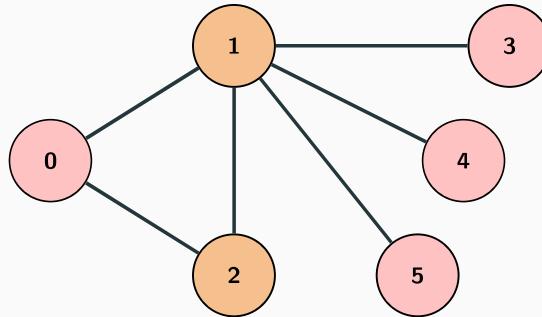
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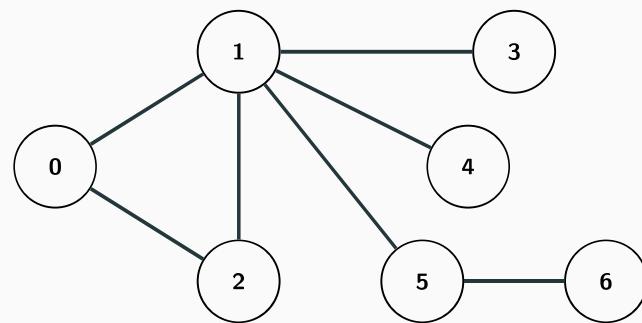
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- The size of a minimum vertex cover plus the size of a maximum independent set is equal to the number of vertices

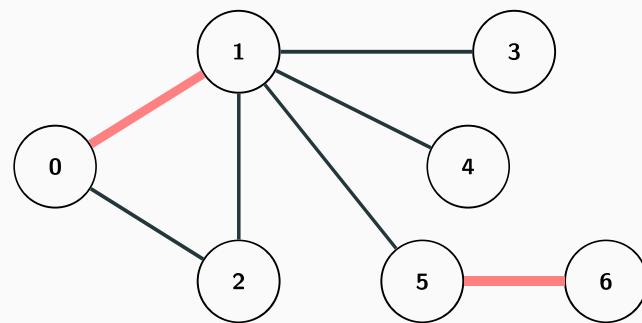
## Maximum matching

- We have an unweighted undirected graph
- A matching is a subset of the edges such that each vertex is adjacent to at most one edge in the subset



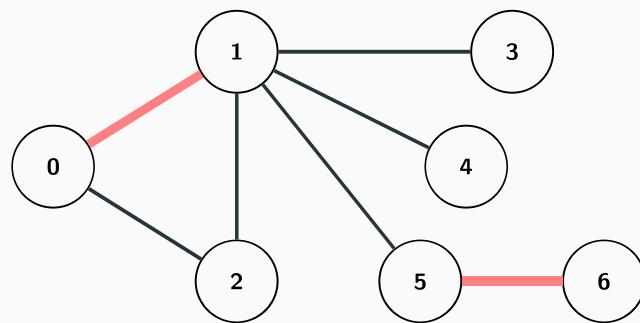
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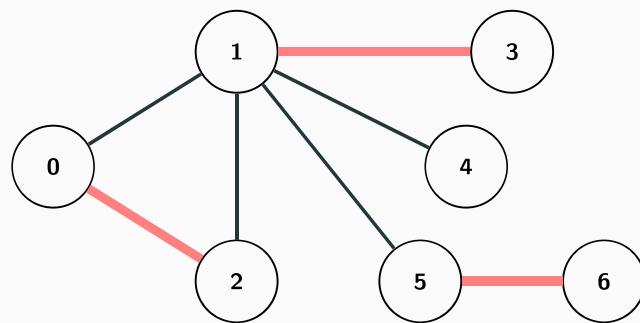
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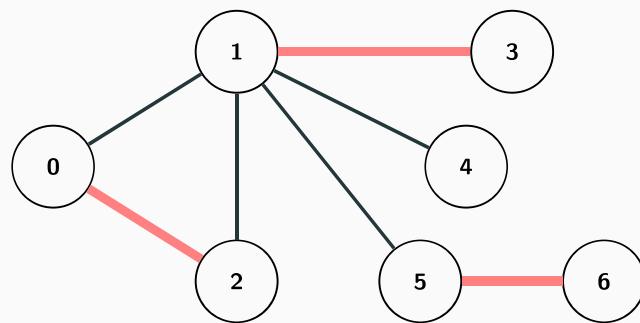
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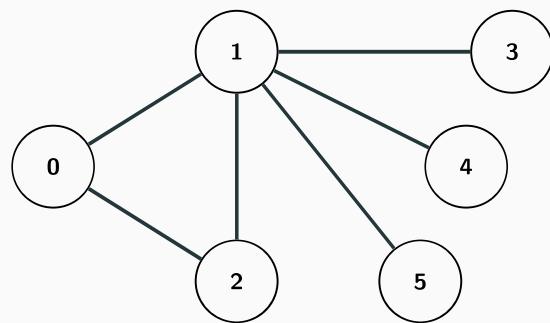
- We have an unweighted undirected graph
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- We want to find a matching of maximum size
- There exists an  $O(V^4)$  algorithm for general graphs, but is pretty complex

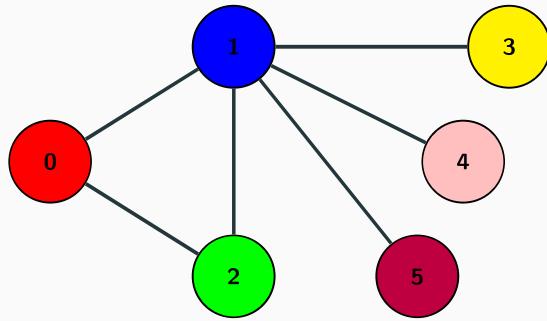
## Graph coloring

- We have an unweighted undirected graph
- A coloring of the graph is an assignment of colors to the vertices such that adjacent vertices have different colors



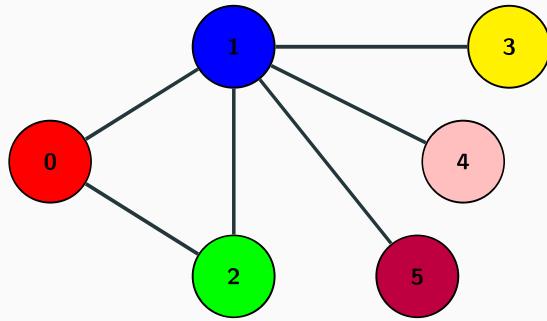
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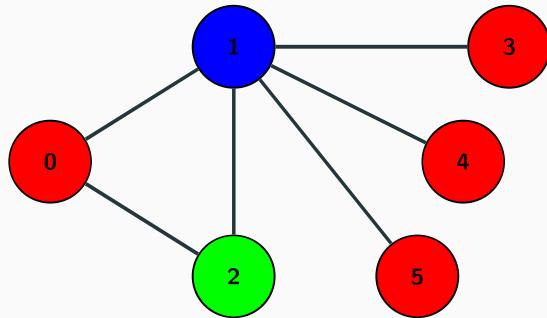
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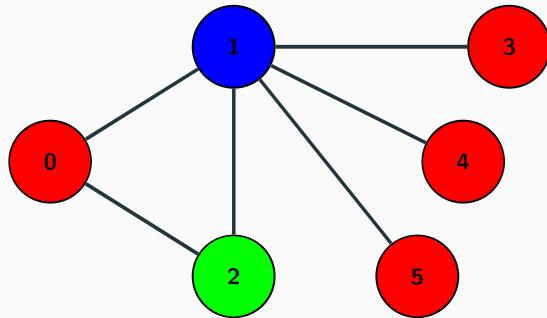
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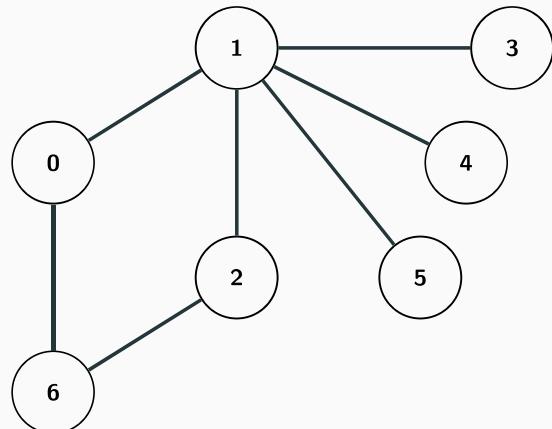
- We want to find a coloring that uses the minimum number of distinct colors
- NP-hard in general graphs

## Special graphs

- All of these problems are hard (in some sense) in general graphs
- But what if we're working with special kinds of graphs?
- Let's look at a few examples

## Bipartite graphs

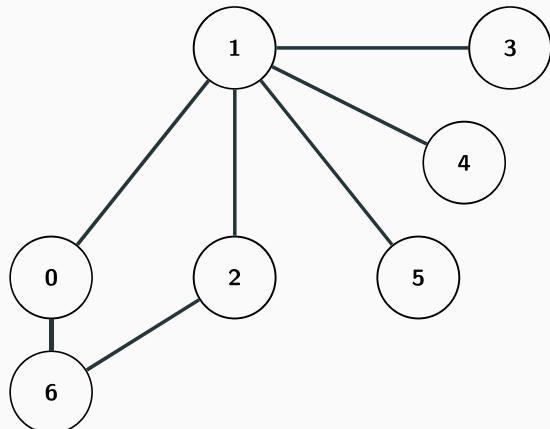
- A graph is bipartite if the vertices can be partitioned into two sets such that for each edge  $(u, v)$   $u$  and  $v$  are in different sets



- How do we check if a graph is bipartite?

## Bipartite graphs

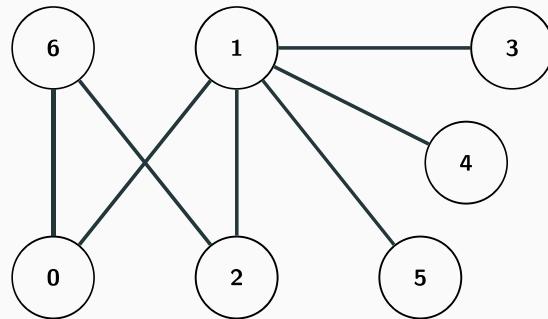
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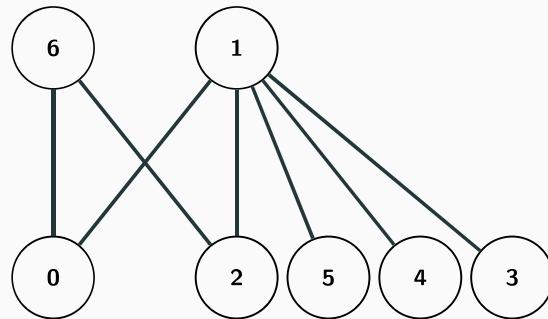
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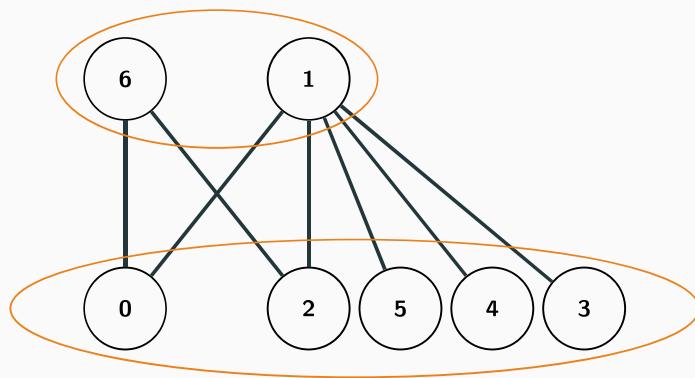
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## Bipartite graphs

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- How do we check if a graph is bipartite?

## Bipartite graphs

- We want to check if we can split the vertices into these two groups
- Take any vertex, and assume that it's in the first group
- Then all of his neighbors must be in the second group
- And then all of their neighbors must be in the first group
- And so on...
- We can do this with a simple depth-first search
- If we ever find a contradiction (i.e. a vertex must both be in the first and second set), then the graph is not bipartite

## Bipartite graphs

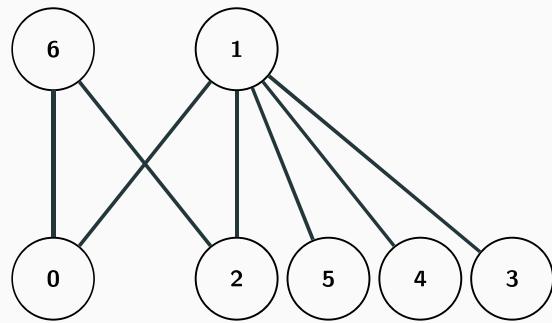
```
vector<int> adj[1000];
vector<int> side(1000, -1);
bool is_bipartite = true;

void check_bipartite(int u) {
    for (int i = 0; i < adj[u].size(); i++) {
        int v = adj[u][i];
        if (side[v] == -1) {
            side[v] = 1 - side[u];
            check_bipartite(v);
        } else if (side[u] == side[v]) {
            is_bipartite = false;
        }
    }
}

for (int u = 0; u < n; u++) {
    if (side[u] == -1) {
        side[u] = 0;
        check_bipartite(u);
    }
}
```

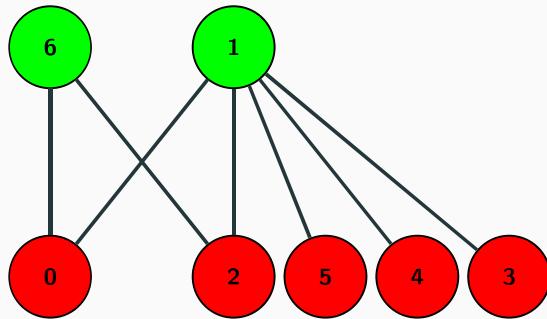
## Coloring bipartite graphs

- What if we want to find the minimum graph coloring of a bipartite graph?



## Coloring bipartite graphs

- What if we want to find the minimum graph coloring of a bipartite graph?



- Simple, one side can be colored with one color, and the second side can be colored with a second color

## Bipartite matching

- Finding a maximum matching in bipartite graphs is very common
  - *see example*
- Soon we'll see an efficient algorithm for finding the maximum matching in a bipartite graph