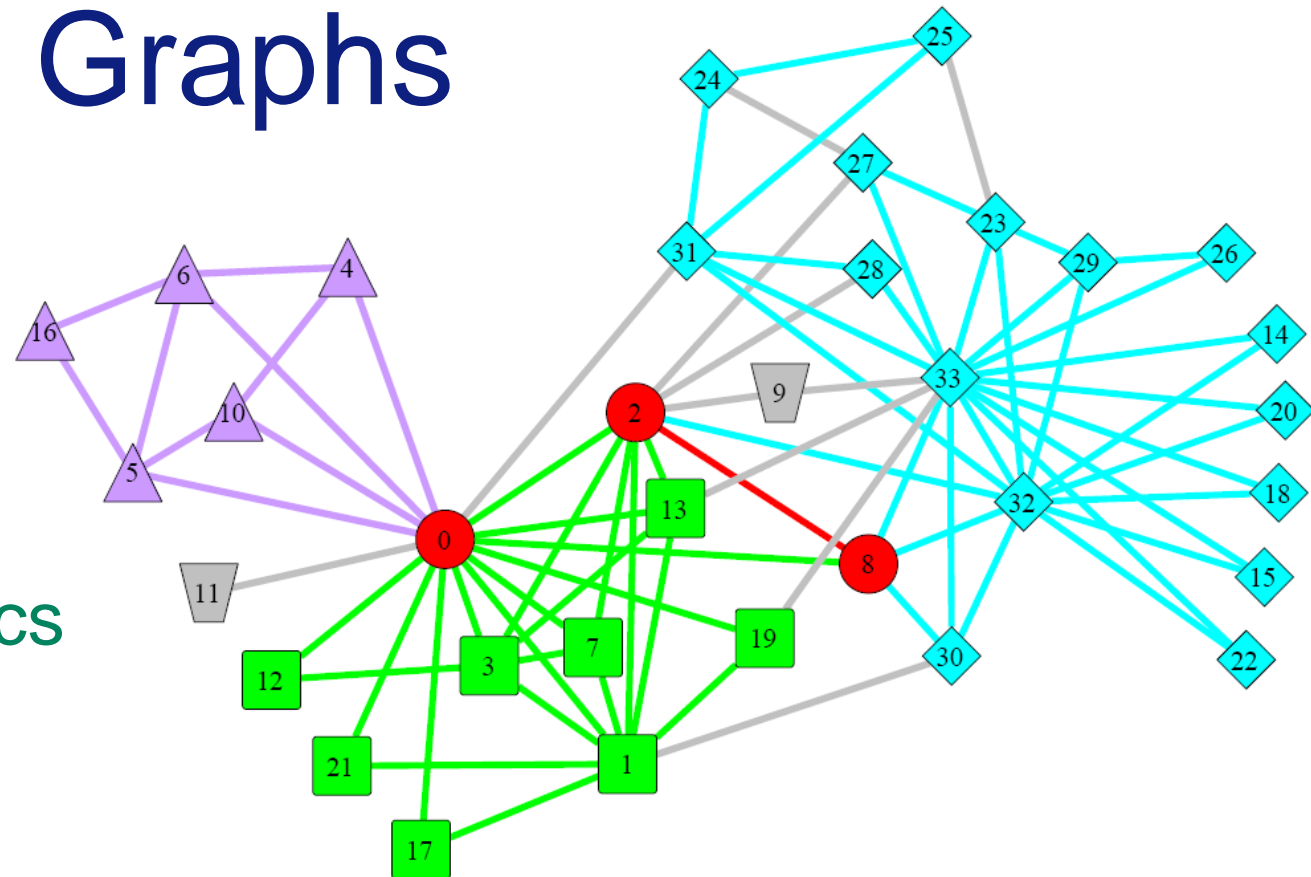


# Clique Graphs



Tim Evans,  
Theoretical Physics  
Imperial College  
London

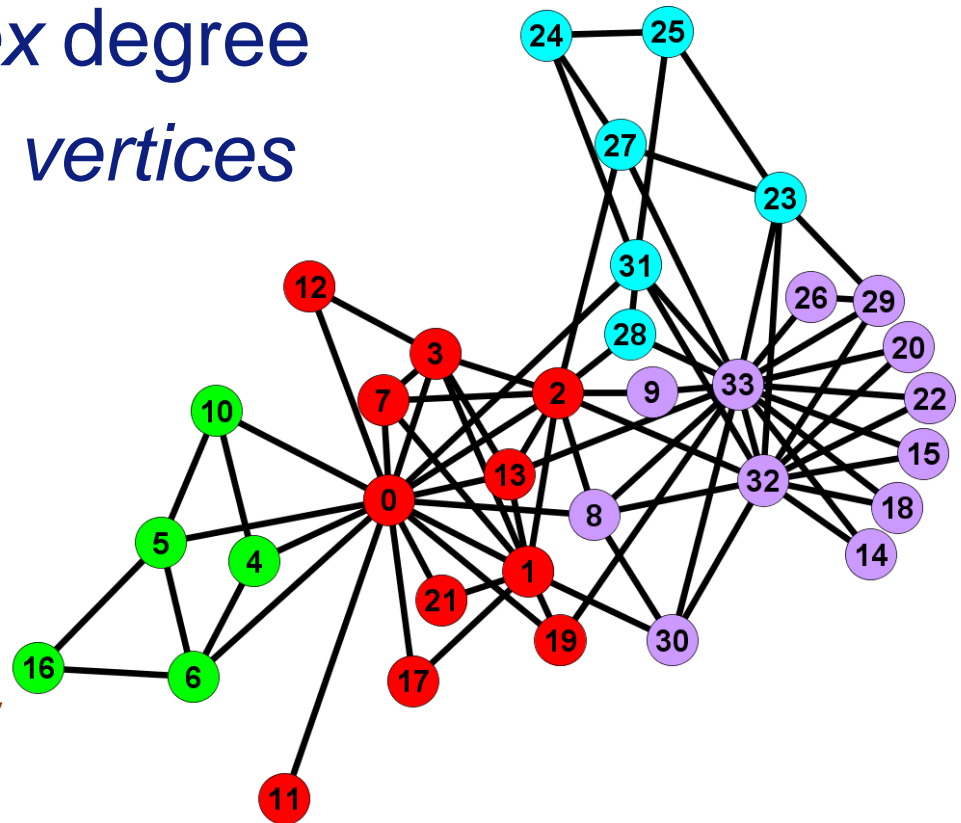
# Vertex Centric Viewpoint

The focus in the literature is often on the vertices:-

- Distributions of *vertex* degree
- Cluster coefficient of *vertices*
- *Vertex* partitions as communities
- .....

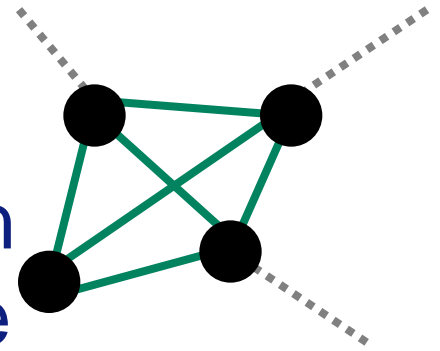
Vertex partition  
of Karate club graph  
with optimal modularity

[Agarwal & Kempe 2007]



# Cliques – complete subgraphs

- Name originates from representation of cliques of people in social science



[Luce and Perry '49]

- Theory of Triadic Closure
- Maximal Clique and Clique Cover problems  
[Bron-Kerbosch algorithm, '73]
- Community detection  
[Palla et al, 05; Yan & Gregory '10]
- Structure Analysis  
[Samudrala & Moulton, 98; Takemoto et al. 07]

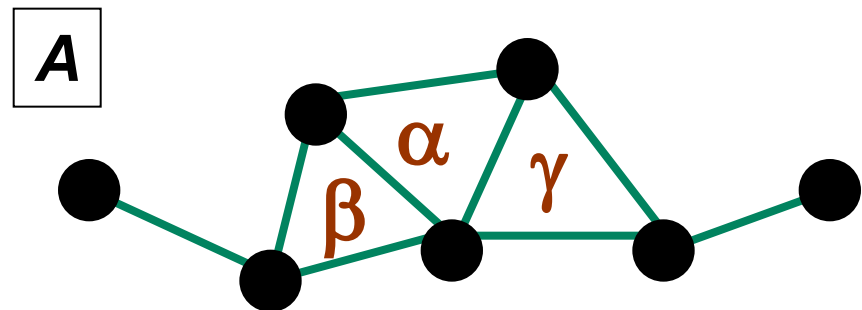
# Give Cliques a Chance – a clique centric viewpoint

Can we shift our view point from vertices to cliques?

Use a **Clique Graph** to represent the way that  $n$ -cliques overlap.

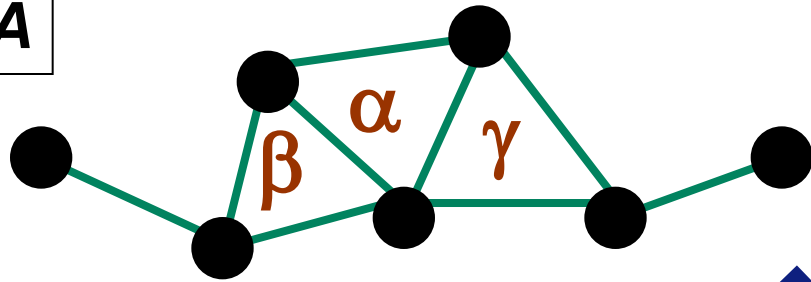
## *Example*

Graph **A** has  
three 3-cliques  
= triangles

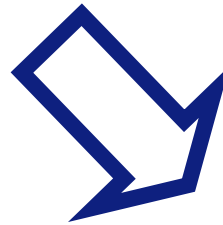


# Clique-Vertex Bipartite Graph Construction

A

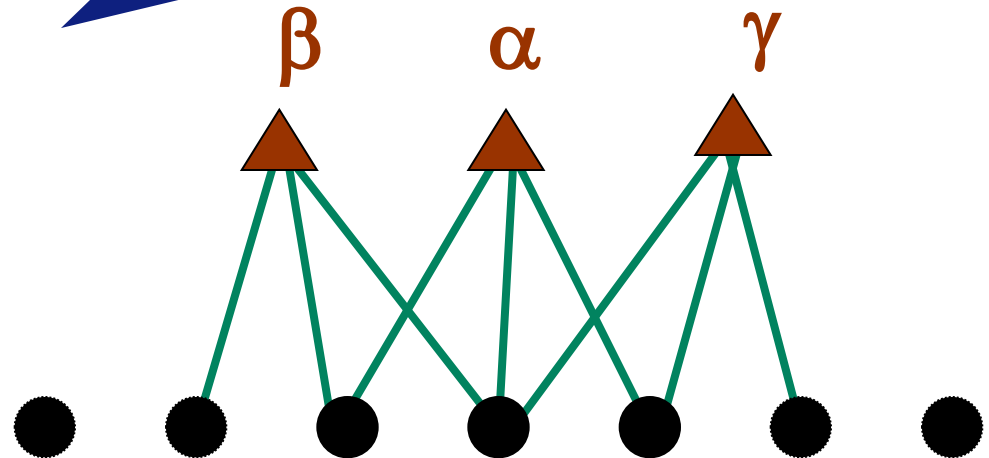


Adjacency matrix  $B_{i\alpha} = 1$   
if vertex  $i$  is in clique  $\alpha$



3-Cliques  $\rightarrow$

Vertices  $\rightarrow$



## Definition of the Basic Clique Graph **C**

Project bipartite graph onto clique vertices.

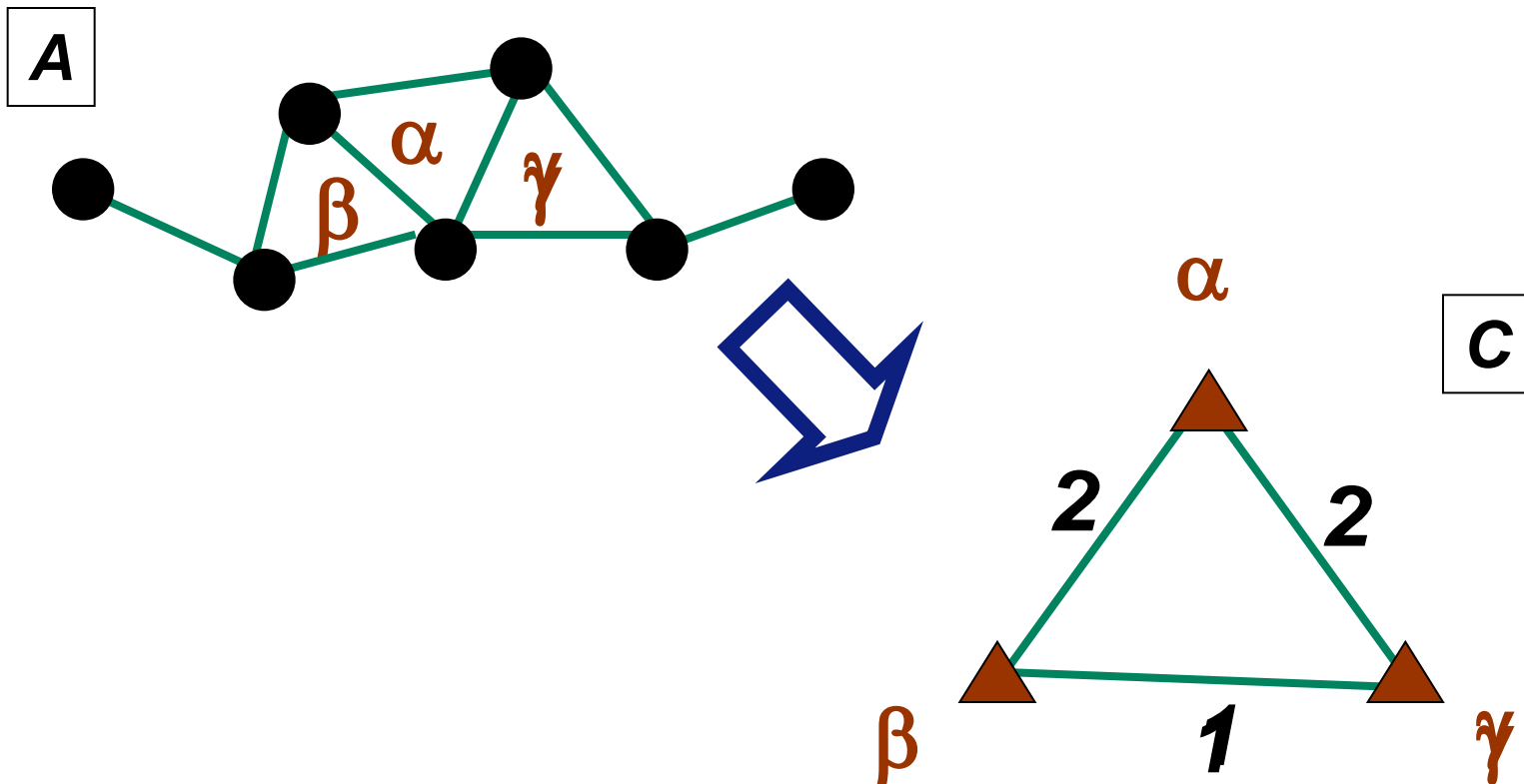
⇒ Weight of an edge in the basic Clique Graph **C** records the number of vertices common to two **n**-cliques in in the original graph **A**.

$$C_{\alpha\beta} = \sum_i B_{i\alpha} B_{i\beta} (1 - \delta_{\alpha\beta})$$

$i = \text{vertex } \mathbf{A}$   
 $\alpha, \beta = \mathbf{n}$ -cliques in  $\mathbf{A}$

## Edge Weights in Basic Clique Graph

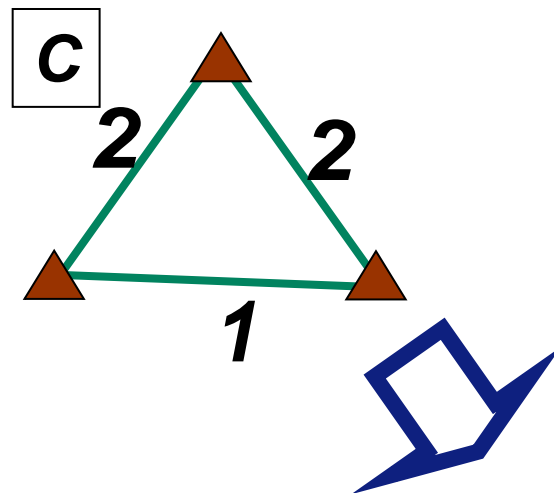
Weight of an edge in the basic Clique Graph **C** records the number of vertices common to two  $n$ -cliques in in the original graph **A**.



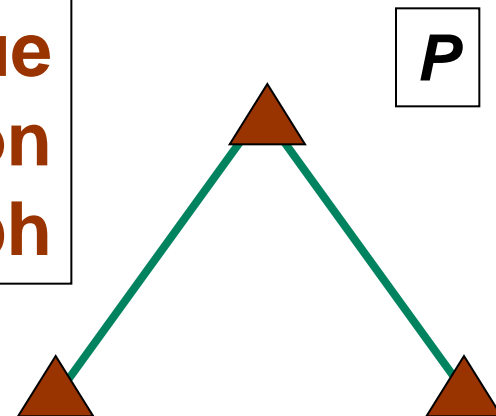
# Clique Percolation Graph Construction

**Threshold edges of  
clique Graph with  $(n-1)$**

i.e. only retain cliques  
which in original graph **A**  
share all but one vertex  
with another clique



**Clique  
Percolation  
Graph**



**Connected Components  
= Communities  
[Palla et al, '05]**

$$P_{\alpha\beta} = \Theta(C_{\alpha\beta} - n + 1)$$

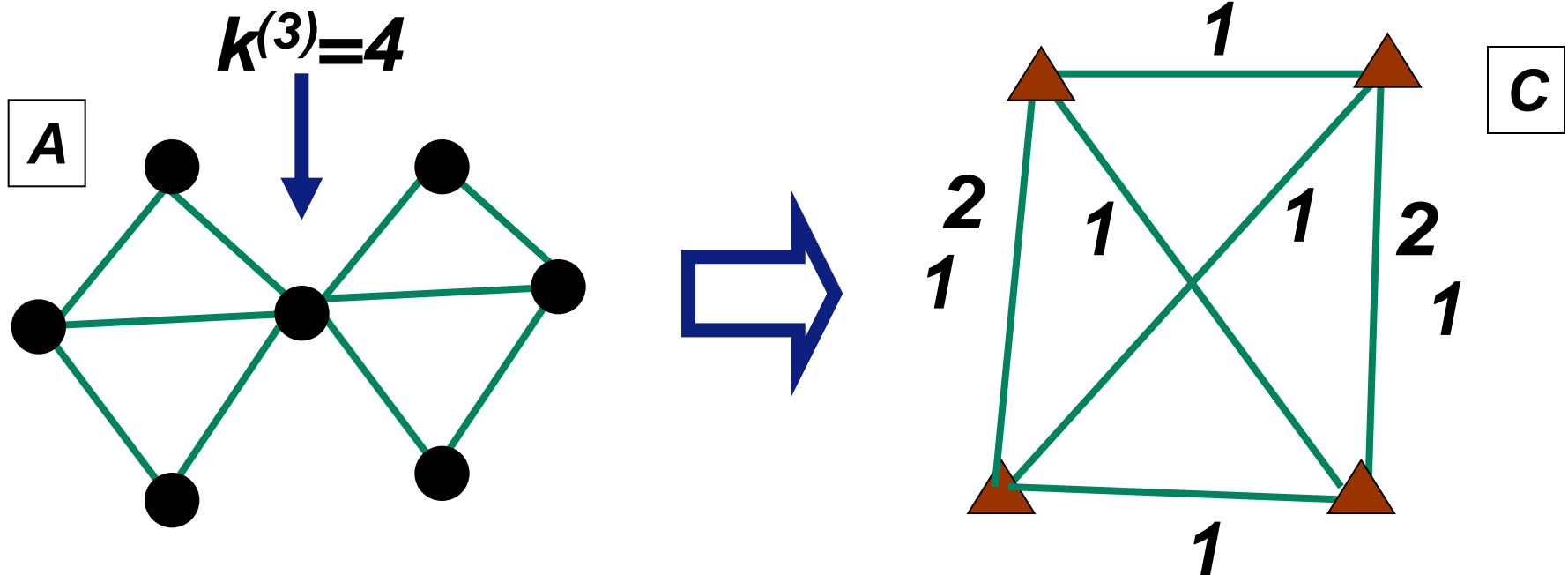


## Basic Clique Graph Weights – A problem

- The weight of basic Clique Graph **C** records the number of vertices common to two **n**-cliques in in the original graph **A**.
- Thus a vertex which is in  $k^{(n)}$  cliques will contribute a total weight of  $k^{(n)} (k^{(n)} - 1)/2$  to the basic clique graph **C**.

# Basic Clique Graph Weights – A problem

**A vertex in a large number of cliques contributes too much weight to the clique graph  $C$ .**



## Better Clique Graph Weight Construction

**A vertex common to a pair of cliques contributes a weight equal to**

$$1/(k^{(n)} - 1)$$

**to an edge in the new weighted Clique Graph  $D$ .**

$$D_{\alpha\beta} = \sum_i \frac{B_{i\alpha} B_{i\beta}}{k^{(n)} - 1} (1 - \delta_{\alpha\beta})$$

(vertex is in  $k^{(n)}$   $n$ -cliques)

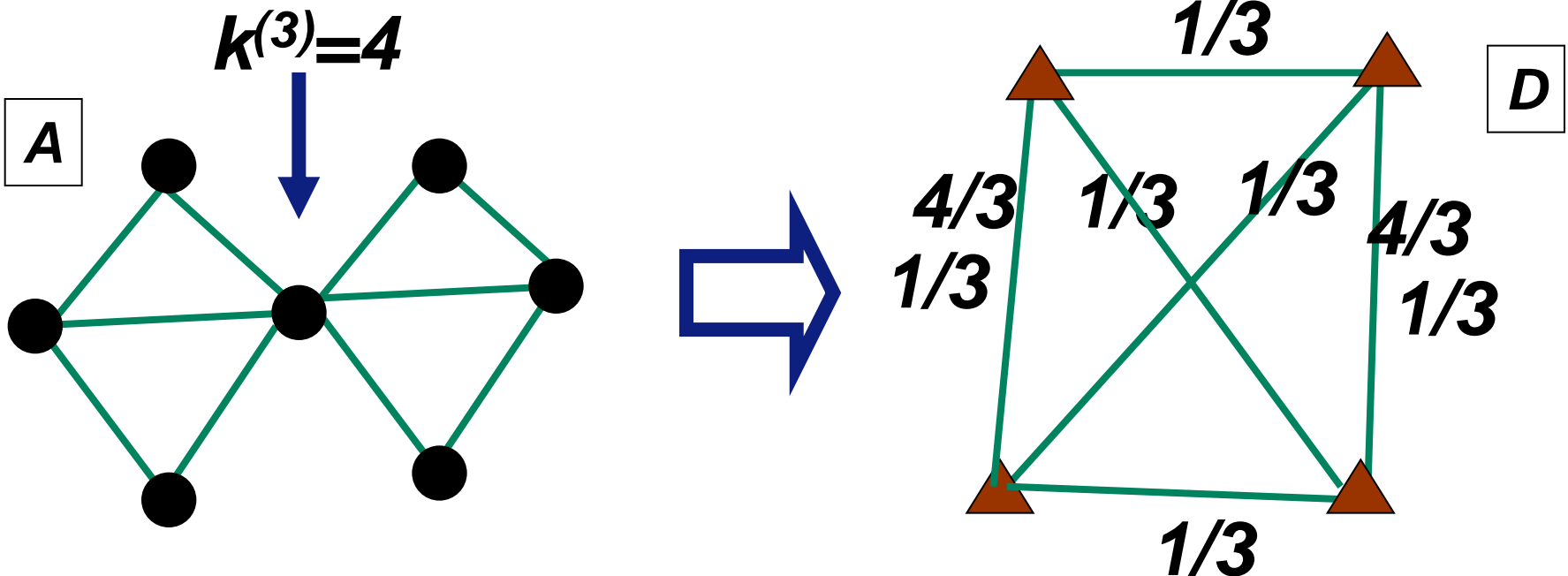
## Better Clique Graph Weight Construction

A vertex common to a pair of cliques contributes a weight equal to

$$1/(k^{(n)} - 1)$$

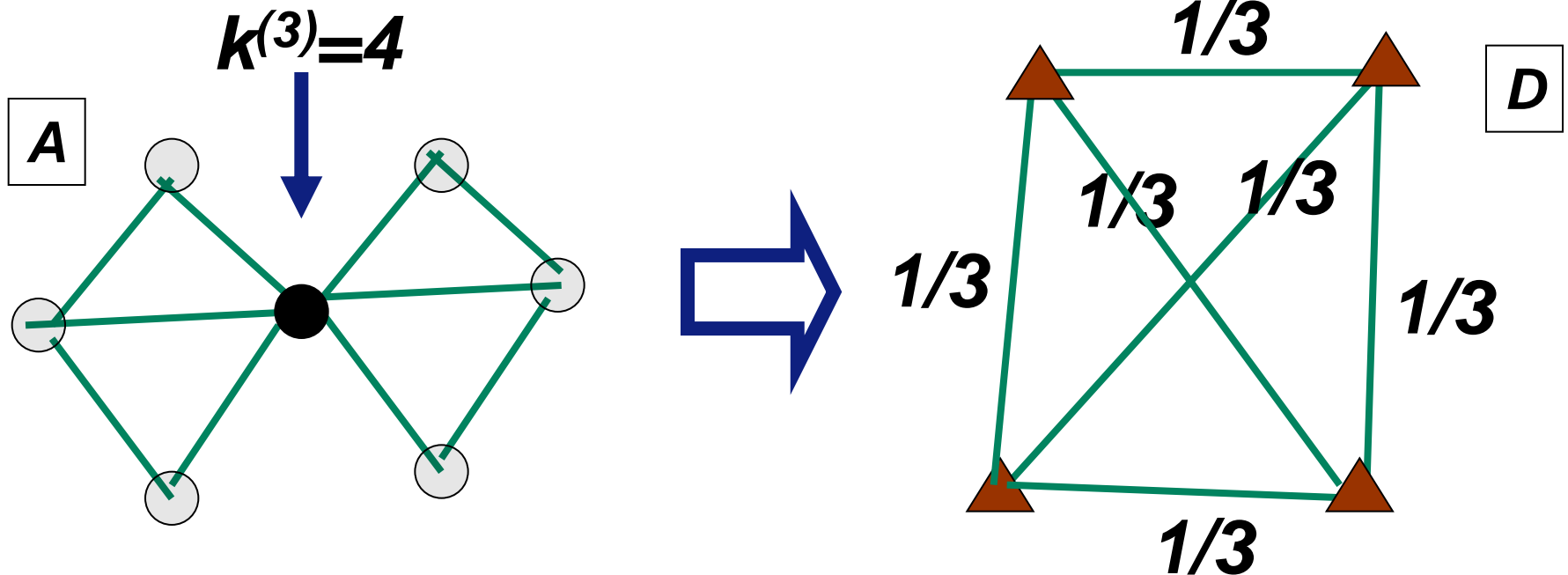
to an edge in the new weighted Clique

Graph *D*. (vertex is in  $k^{(n)}$   $n$ -cliques)



## Better Clique Graph Weight Construction

Now each vertex  $i$  in the original graph  $A$  contributes a weight to the new weighted Clique Graph  $D$  proportional to the number of cliques containing that vertex,  $k^{(n)}$ .



# Applications

Now apply **ANY** standard vertex based algorithm to vertices of clique graph **C** or **D** or other weighted version

- Degree distribution  $\Rightarrow$  clique overlap distribution
- Cluster coefficient  $\Rightarrow$  overlap of 3 cliques
- Community detection  $\Rightarrow$  clique communities
- Etc, etc

# Clique Graphs for Community Detection

- Produce Clique graph
  - Use favourite vertex partition method in clique graph
  - Hence assign each clique to a single community
  - Deduce community membership of vertices and edges of original graph from clique membership.
- ⇒ Thus vertices and edges maybe part of more than one community.
- ⇒ Overlapping Communities.

# Cliques in Karate Club

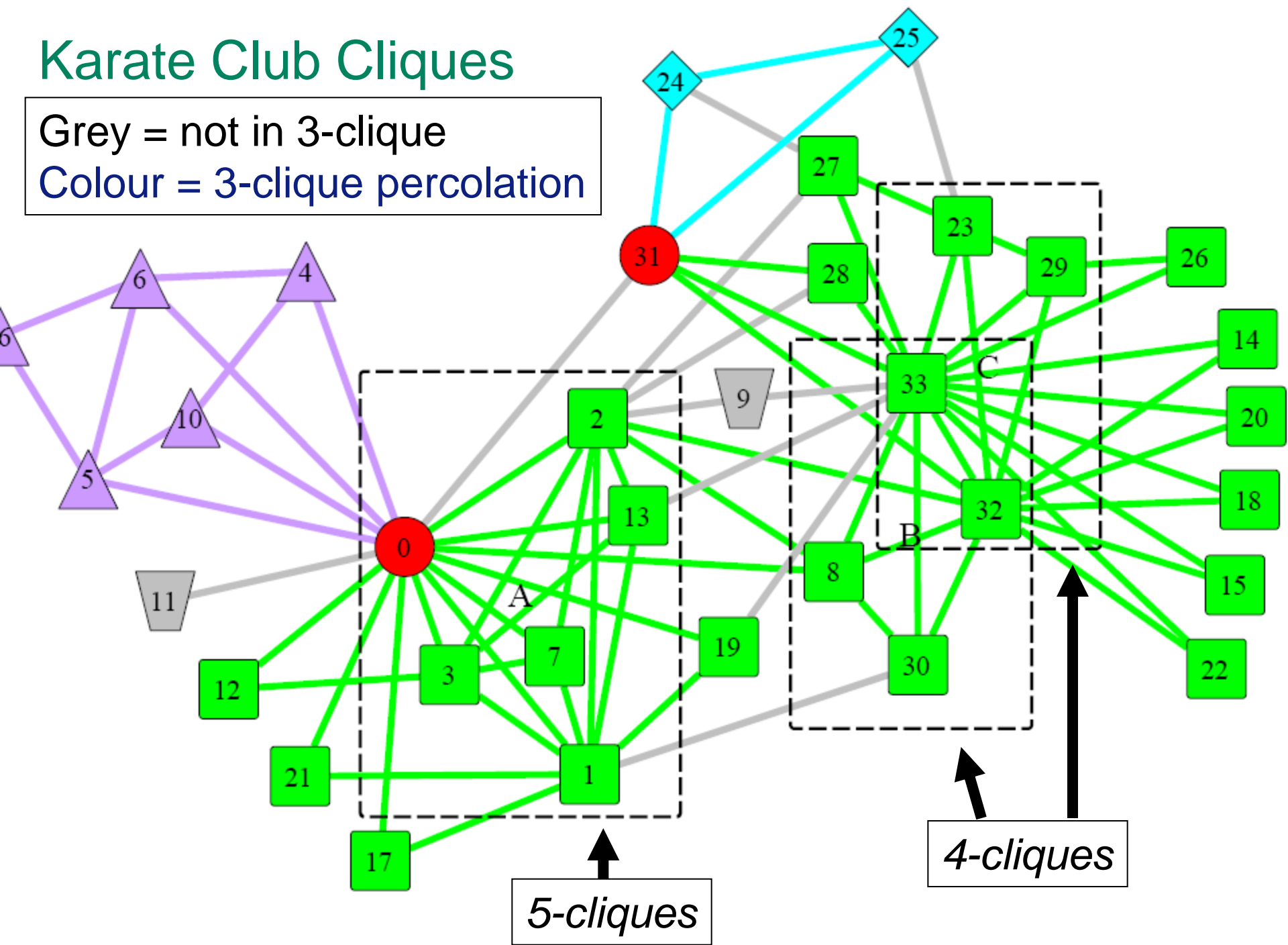
- Almost everything is in a 3-clique
  - just 2 vertices and 9 edges not in 3-cliques
- One group of 6 vertices centred on Instructor have two percolating 5-cliques
- Two other 4-cliques centred on chief officers, non-percolating



# Karate Club Cliques

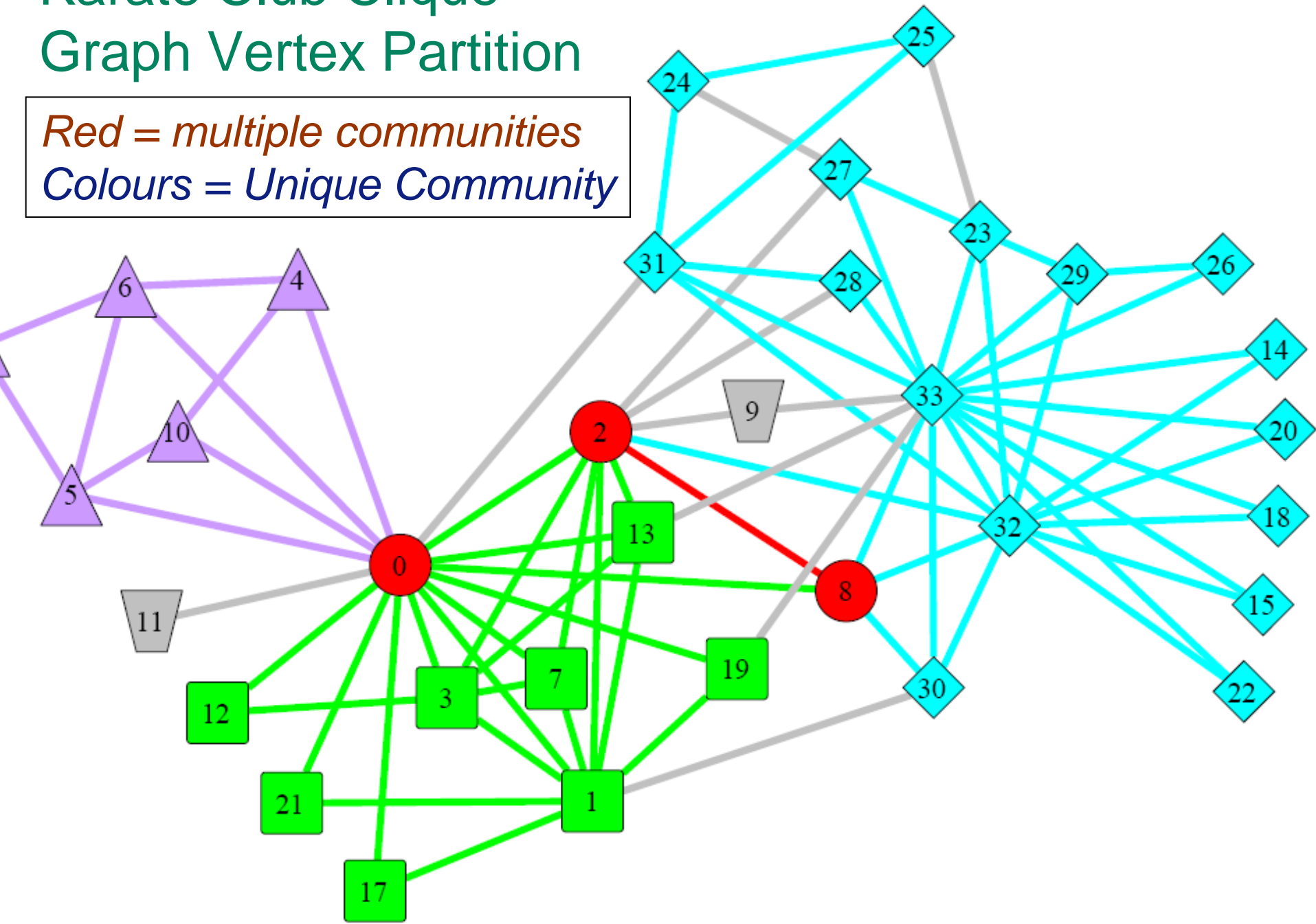
Grey = not in 3-clique

Colour = 3-clique percolation



# Karate Club Clique Graph Vertex Partition

*Red = multiple communities*  
*Colours = Unique Community*



# Analysis of Karate Club Results

- Clique percolation doesn't work here
- Despite the fact that almost everything is in a 3-clique, vertex partitioning of a clique graph works extremely well.

## Conclusions

- Clique graphs move focus from vertices to cliques with minimal effort
- Altering weights avoids problem of over representation of high degree vertices
- 2-clique graphs are *Weighted Line Graphs*  
[Evans and Lambiotte, 09]
- Generalisation to motifs straightforward
- Community detection on clique graph produces overlapping vertex communities for original graph