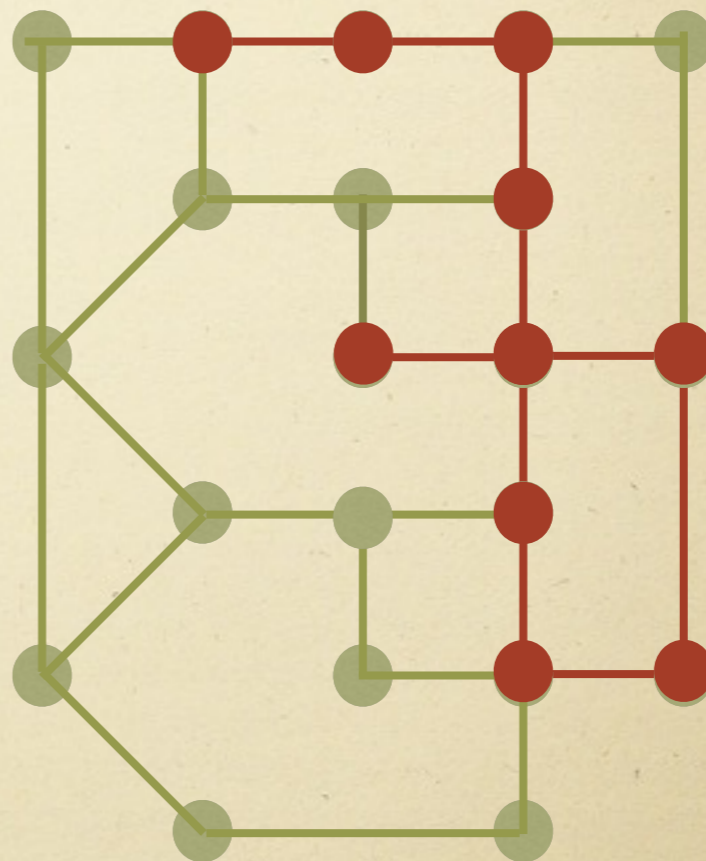


Graph homomorphism and similarity measures

Eunjung Kim, LAMSADE/CNRS
ESIGMA Kick-off meeting, 31 May 2018

Walk



Graph kernel method

graphs





mapping ϕ 

(feature) vectors

[2,3,5,5,6,3,2,0,4]

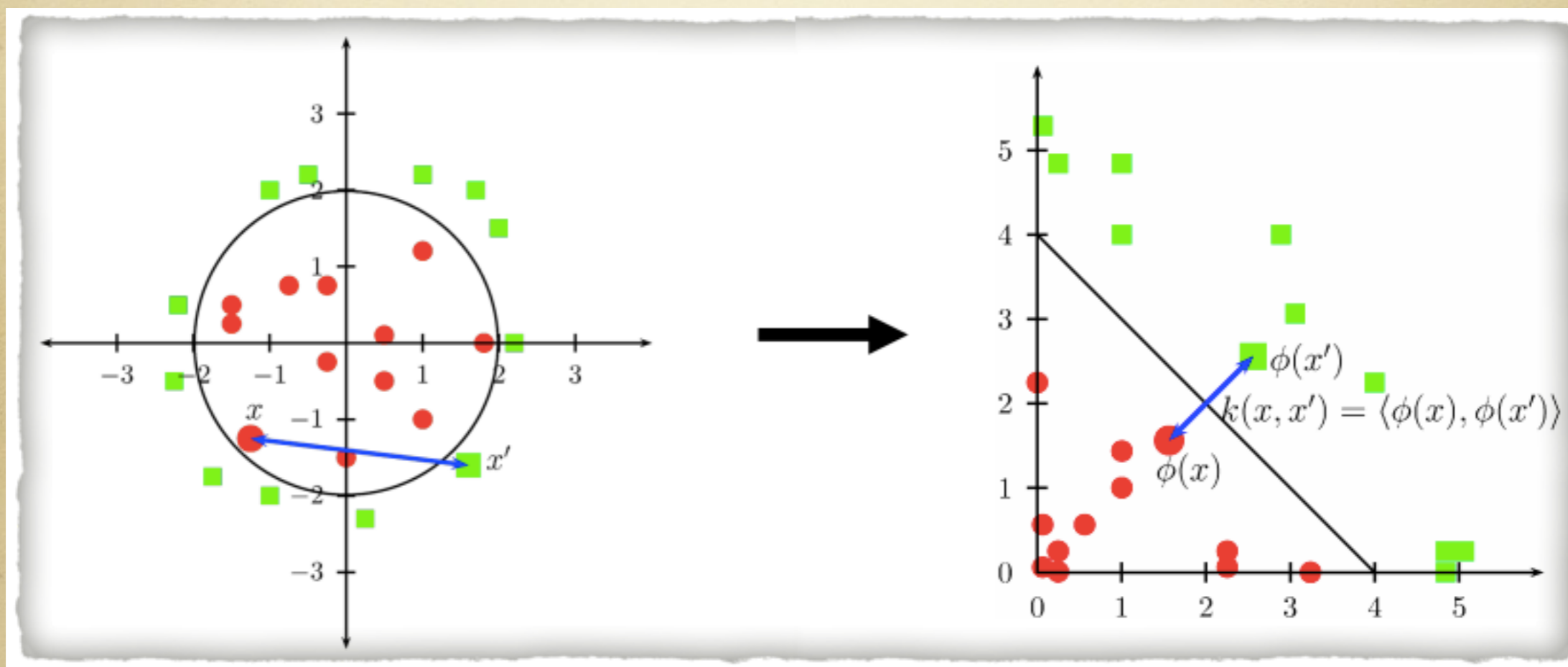
[1,3,4,2,5,3,4,0,0]

dist???(, )

$k(G,H) = \langle \phi(G), \phi(H) \rangle$

* kernel = inner product on some space,
which in turn defines a metric via $d(x,y) := \sqrt{\langle x-y, x-y \rangle}$.

- Easy computation: we can avoid actually defining an explicit mapping, computing the mapping, or doing coordinate-wise computation in the feature space.
- Good math make it work: if a function k meets some criteria (p.d.kernel), then it's an inner product of some space and a distance is well-defined. ()

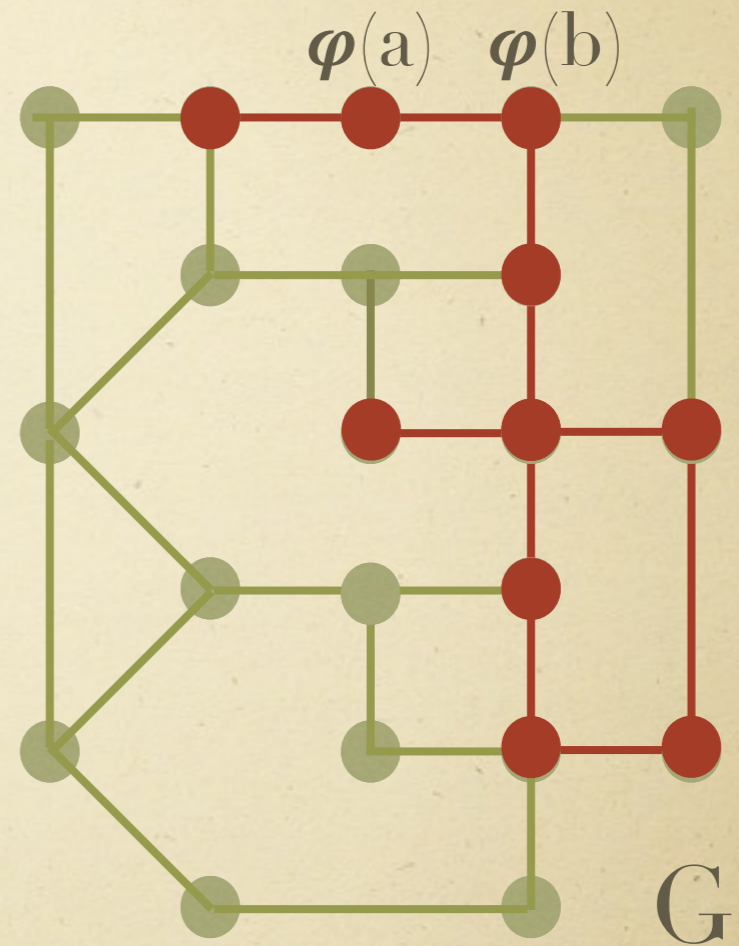


[Borgwardt and Stegle's slide "An introduction to Graph Kernels" 2010]

Graph kernel (via random walk)

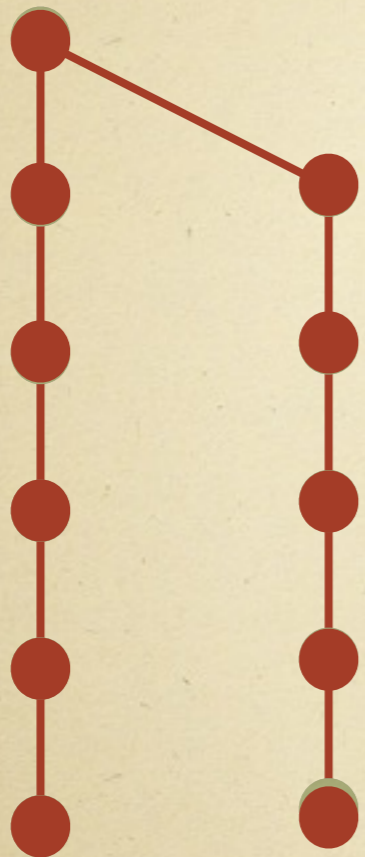
Idea, as much as I get from Ioannis' talk:

Homomorphism

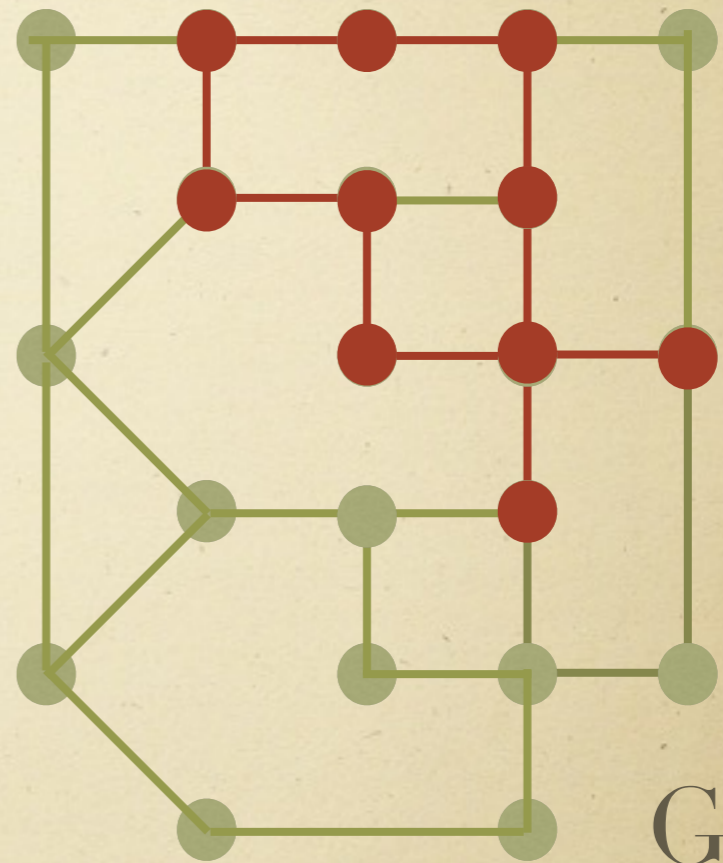


mapping $\varphi: V(F) \rightarrow V(G)$
s.t. (a,b) implies $(\varphi(a), \varphi(b))$

Homomorphism



F



G

mapping $\varphi: V(F) \rightarrow V(G)$ s.t. (a,b) implies $(\varphi(a), \varphi(b))$

Isomorphism via homomorphism

G and H are isomorphic
if and only if

$\#HOM(F,G) = \#HOM(F,H)$ for every graph F

László Lavász, “Operations with structures”
. Acta Mathematica Hungarica 1967

Homomorphism vector

G

$$\text{HOM}(G) = [\# \text{HOM}(F1, G), \# \text{HOM}(F2, G), \# \text{HOM}(F3, G), \dots]$$

H

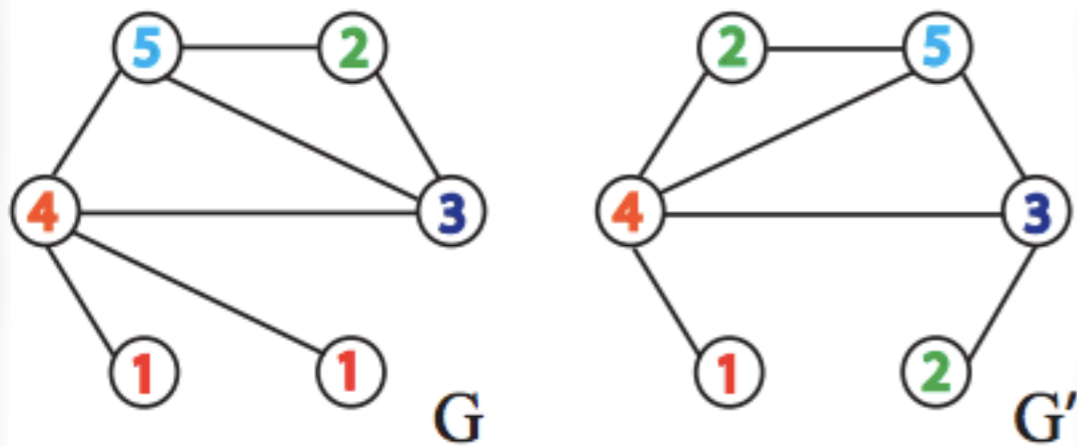
$$\text{HOM}(H) = [\# \text{HOM}(F1, H), \# \text{HOM}(F2, H), \# \text{HOM}(F3, H), \dots]$$

\neq

G and H are isomorphic
if and only if
 $\text{HOM}(F, G) = \text{HOM}(F, H)$ for every graph F

Color-refinement: heuristic isomorphism test

Given labeled graphs G and G'

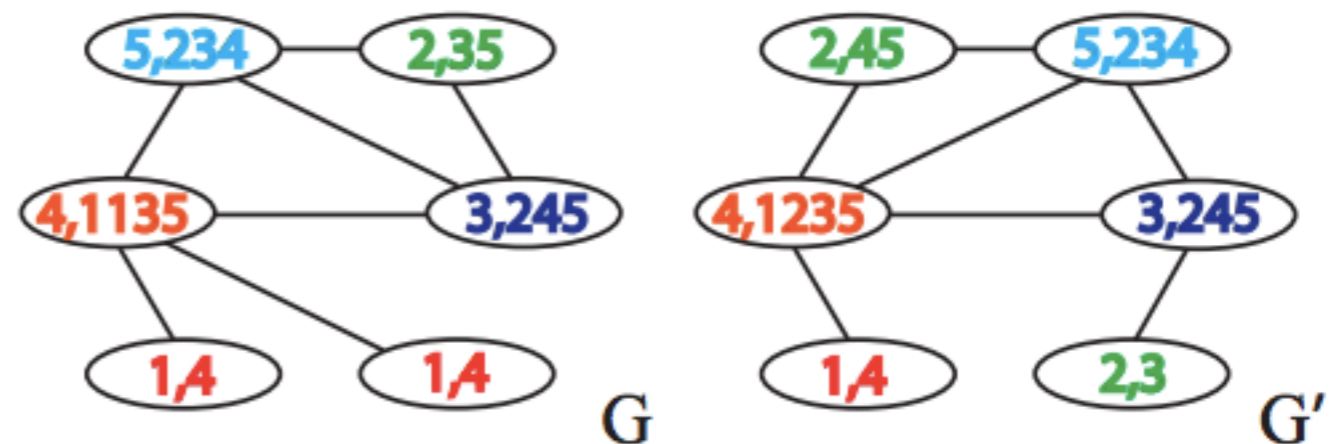


a

Figures excerpted from
Shervashidze et. al. 2011

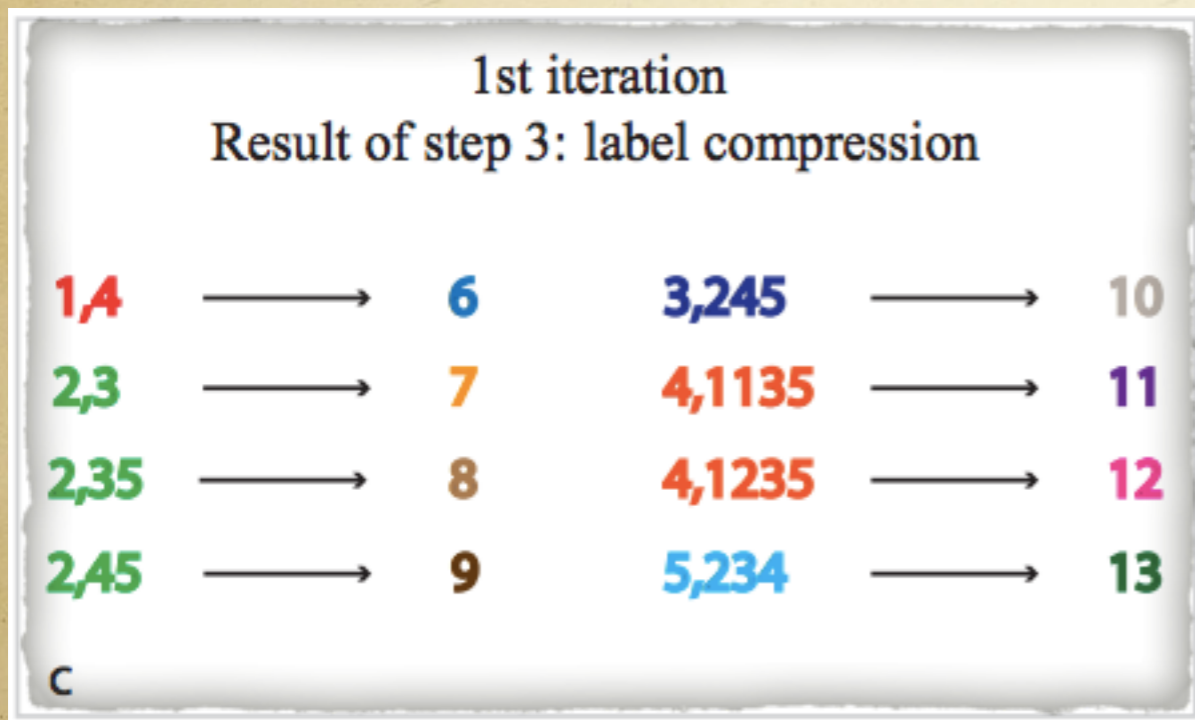
1st iteration

Result of steps 1 and 2: multiset-label determination and sorting

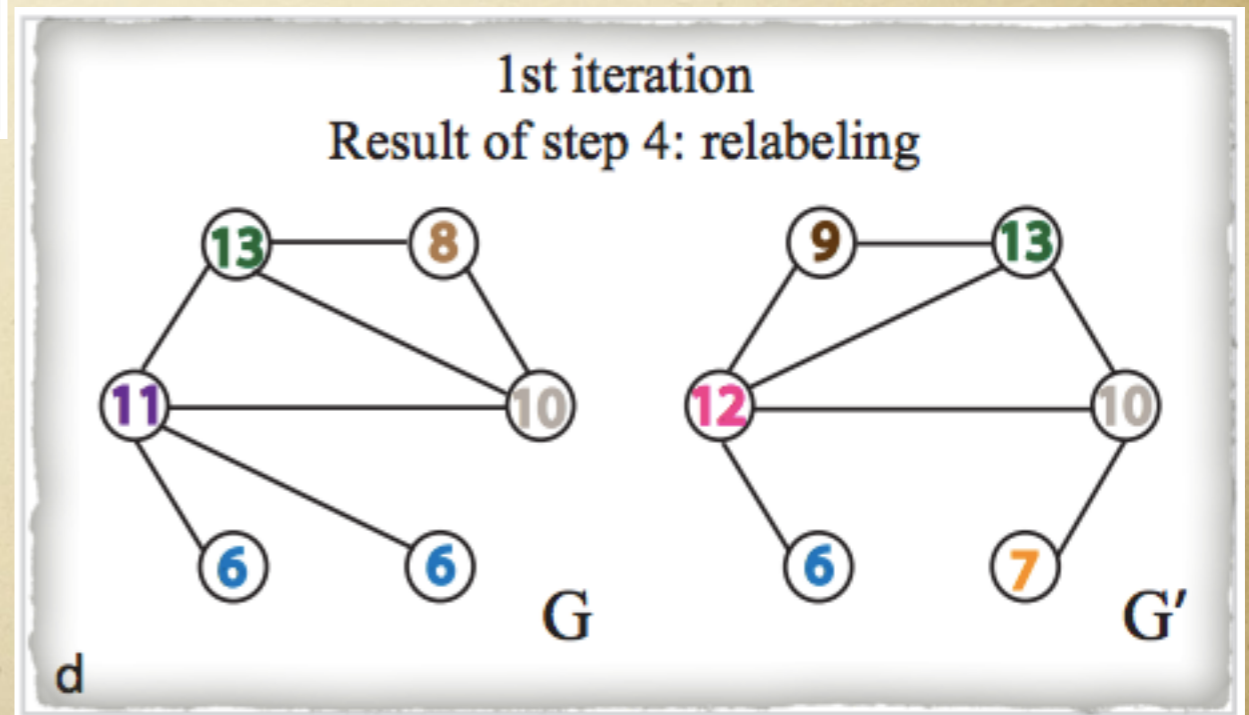


b

Color-refinement

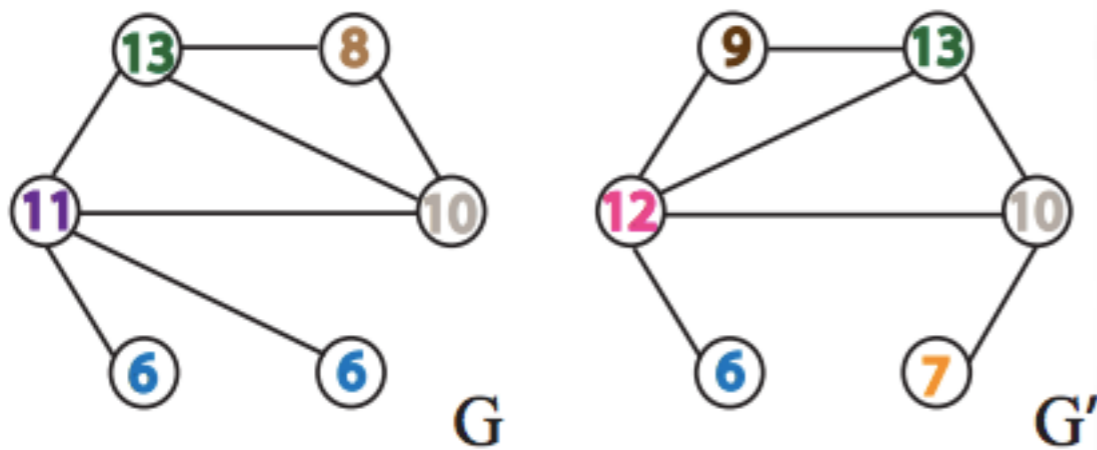


Figures excerpted from
Shervashidze et. al. 2011



Color-refinement

1st iteration
Result of step 4: relabeling



d

Figures excerpted from
Shervashidze et. al. 2011

End of the 1st iteration
Feature vector representations of G and G'

$$\varphi_{WLsubtree}^{(1)}(G) = (2, 1, 1, 1, 1, 2, 0, 1, 0, 1, 1, 0, 1)$$

$$\varphi_{WLsubtree}^{(1)}(G') = (1, 2, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1)$$

Counts of
original
node labels

Counts of
compressed
node labels

$$k_{WLsubtree}^{(1)}(G, G') = \langle \varphi_{WLsubtree}^{(1)}(G), \varphi_{WLsubtree}^{(1)}(G') \rangle = 11.$$

e

Color-refinement

color-refinement
algorithm

ultimate
isomorphism test

G

$\text{HOM}(G) | \text{all graphs}$

What is this vector?

#

H

$\text{HOM}(H) | \text{all graphs}$

Indistinguishable under color-refinement

G and H are indistinguishable under color-
refinement
if and only if
 $\text{HOM}(F,G) = \text{HOM}(F,H)$ for every tree F

Dell, Grohe and Rattan, “Lovász Meets Weisfeiler and Leman”

ICALP 2018

Color-refinement from Lovász' perspective

color-refinement
algorithm

ultimate
isomorphism test

G

$\text{HOM}(G)|_{\text{all trees}}$

$\text{HOM}(G)|_{\text{all graphs}}$

||

#

H

$\text{HOM}(H)|_{\text{all trees}}$

$\text{HOM}(H)|_{\text{all graphs}}$

Color-refinement from Lovász' perspective

1-Weisfeiler-Leman
algorithm
(color-refinement)

k-Weisfeiler-Leman
algorithm

n-Weisfeiler-Leman
algorithm

G

HOM(G)|all trees

HOM(G)| $tw \leq k$

HOM(G)|all graphs

#

H

HOM(H)|all trees

HOM(H)| $tw \leq k$

HOM(H)|all graphs

Take-home message

“State-of-the-art graph kernel methods
work for a good reason, now with
theoretical support.”

“How can we use it?”

Future Work?