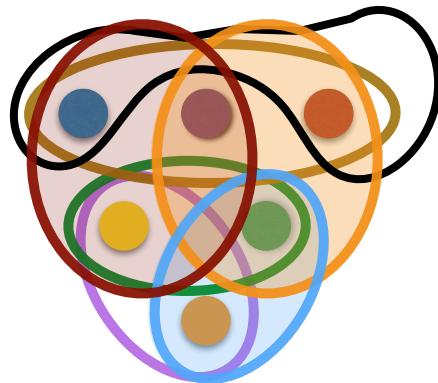
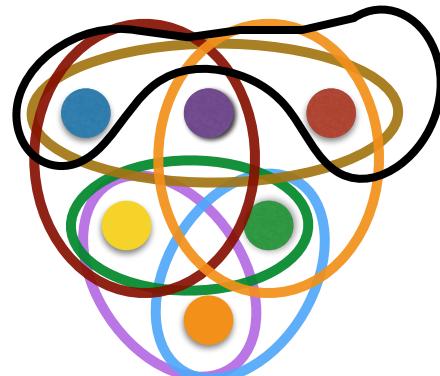


# Set cover, linear programming and randomized rounding



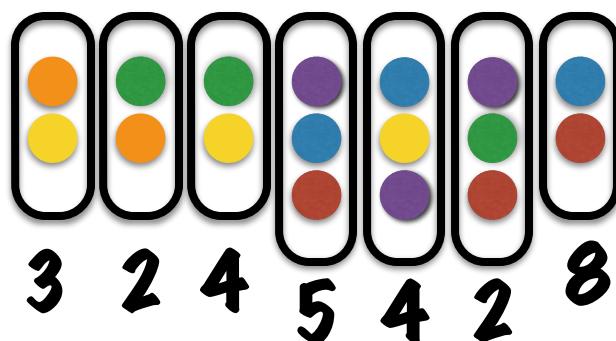
# The set cover problem



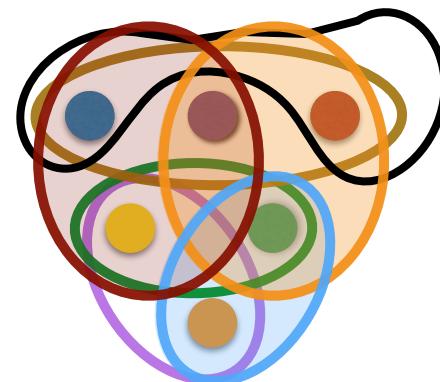
Elements



Subsets with costs



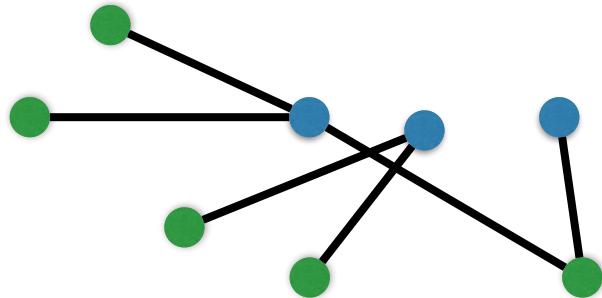
Choose subsets  
Cover elements  
at min cost



$$\text{Cost} = 4 + 2 + 2 = 8$$

Does this ring a bell?

# Vertex cover



Cover all edges  
with the fewest  
vertices

edges = elements  
vertices = sets

Integer program  
for  
Set cover

**Variable for subset S:**

$$x_S = 1$$

**iff S in cover**

**Constraint for element i:**

$$\sum_{S: e \in S} x_S \geq 1$$

**Objective:**

$$\min \sum_S c_S x_S$$

# Linear programming relaxation

$$\min \sum_S c_S x_S$$

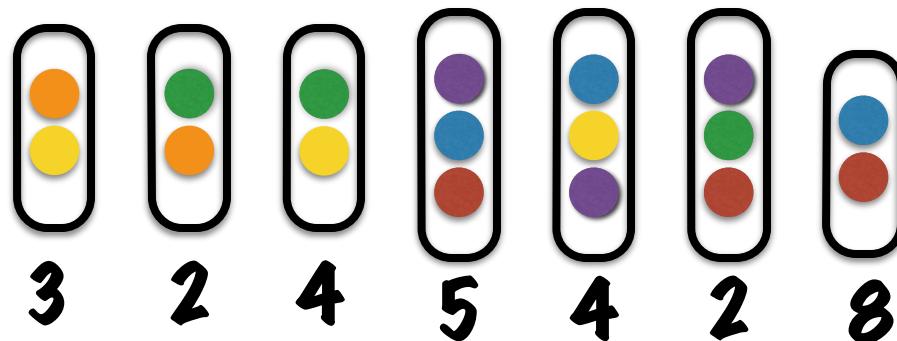
such that

$$\begin{cases} \sum_{S: e \in S} x_S \geq 1 & \forall e \\ 0 \leq x_S \leq 1 & \forall S \end{cases}$$

# Rounding

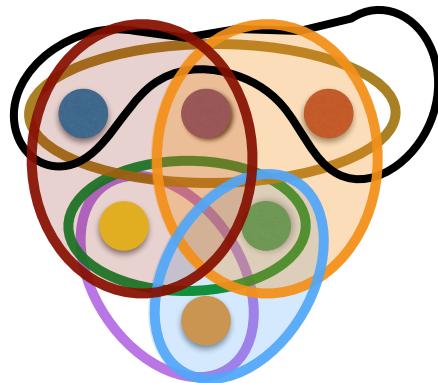
Like vertex cover:  
round to 1 iff  $x_u \geq 1/2$

Fails

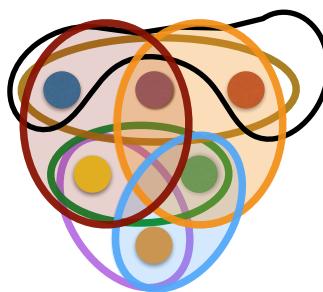


$$x_S : \frac{1}{3}, \frac{2}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{2}, \frac{1}{3}$$

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



## Linear programming relaxation

$$\min \sum_S c_S x_S$$

such that

$$\begin{cases} \sum_{S:e \in S} x_S \geq 1 & \forall e \\ 0 \leq x_S \leq 1 & \forall S \end{cases}$$

How do we round the LP solution?

# Randomized Rounding: An algorithm

$x_i = .9$  : should probably go to 1

$x_i = .1$  : should probably go to 0

Cannot fix a threshold

Idea: randomized rounding

$x_i = .8 \implies$  round to 1  
w.p. 80%

# New rounding algorithm

For each set  $S$   
with probability  $x_S$   
put  $S$  in the cover

# **Analysis**

**Is it efficient?  
Is the output a cover?  
How good is it?**

# **Analysis**

**Is it efficient?  
Yes**

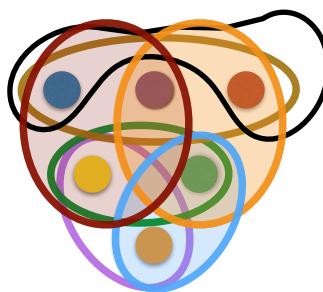
# **Analysis**

**Is the output a cover?  
Maybe, maybe not**

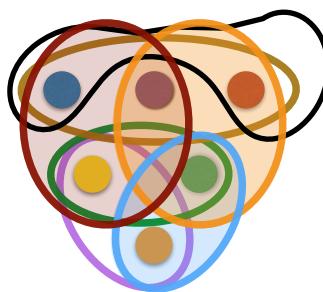
# Analysis

**How good is it?**  
*It depends...*

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



**How good is it?**  
**It depends...**

$$\text{Value}(\text{Output}) = \sum_{S \text{ in cover}} c_S$$

**How good is it *on average*?**

$$E[\sum_S 1(S \text{ in cover})c_S] = ?$$

# Linearity of expectation

$$E[A + B] = E[A] + E[B]$$

$$\begin{aligned} E\left[\sum_S 1(S \text{ in cover})c_S\right] &= \\ \sum_S E[1(S \text{ in cover})c_S] \end{aligned}$$

$$E[\lambda X] = \lambda E[X]$$

$$\begin{aligned} \sum_S E[1(S \text{ in cover}) c_S] &= \\ \sum_S E[1(S \text{ in cover})] c_S \end{aligned}$$

$$\begin{aligned} E[1(S \text{ in } \text{cover})] &= \\ \Pr[S \text{ in } \text{cover})] \end{aligned}$$

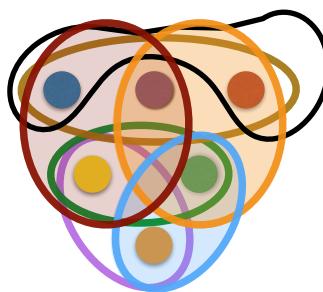
$$\Pr[\mathbf{S} \text{ in cover}] = \mathbf{x_S}$$

**Together**

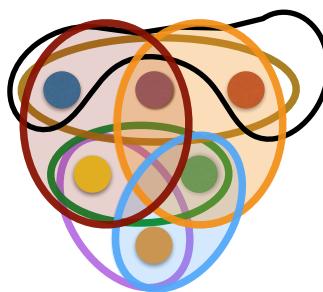
$$E[Value(Output)] = \sum_S x_S c_S$$

**Value of the linear program!**

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



**Is the output a cover?  
Maybe, maybe not**

**Number of elements covered:**

$$\sum_e 1(e \text{ covered})$$

**On average:**

$$\sum_e \Pr[e \text{ covered}]$$

**Consider an element  $e$ .  
With what probability  
is it covered by output?**

$\Pr[\mathbf{e \text{ covered}}] =$   
 $\Pr[\text{there exists } S \text{ in output:}$   
 $\mathbf{e \in S}]$

$\Pr[\text{there exists } S \text{ in output:}$   
 $e \in S] =$   
 $1 - \Pr[\text{for all } S \text{ containing } e:$   
 $S \text{ not in output}]$

# Independence

If independence:

$$\Pr[A \text{ and } B] = \Pr[A] \times \Pr[B]$$

$\Pr[\text{for all } S \text{ containing } e:$

$S \text{ not in output}] =$

$$\prod_{S:e \in S} \Pr[S \text{ not in output}]$$

$$\Pr[S \text{ not in output}] = 1 - x_S$$

# Together

$$\Pr[\text{e covered}] = 1 - \prod_{S:e \in S} (1 - x_S)$$

# Algebra

$$X = e^{\ln X}$$

$$\ln(XY) = \ln X + \ln Y$$

$$\prod_{S:e \in S} (1 - x_S) = e^{\sum_{S:e \in S} \ln(1 - x_S)}$$

# Algebra

$$\ln(1 - X) \leq -X$$

$$e^{\sum_{S:e \in S} \ln(1-x_S)} \leq e^{-\sum_{S:e \in S} x_S}$$

# Use LP constraint

$$\sum_{S:e \in S} x_S \geq 1$$

$$e^{-\sum_{S:e \in S} x_S} \leq e^{-1}$$

# Combining

$$\Pr[e \text{ covered}] \geq 1 - 1/e$$

**Average number of  
elements covered:**

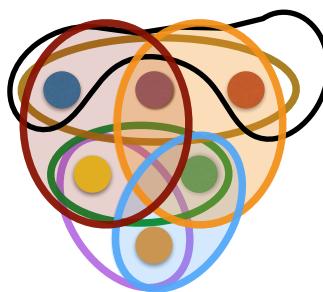
$$\#(\text{elements})(1 - 1/e)$$

$$1 - 1/e = 0.63\dots$$

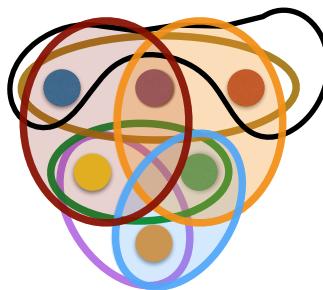
# Recap

**Randomized rounding gives  
collection of sets  
with average cost  
at most OPT  
and covering on average  
63% of the elements.**

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



# Getting a set cover

Idea: repeat!

# Randomized rounding algorithm

$n = \# \text{elements}$

Repeat  $\ln(n) + 3$  times

For each  $S$

Put  $S$  in cover w.pr.  $x(S)$   
(if not there already)

Note:  $e^3 = 20.0\dots$

# Cost

In expectation:  
at most  
 $(\ln(n) + 3)OPT$

# Correctness

$$\Pr[\text{cover}] = 1 - \Pr[\text{not cover}]$$

$$\begin{aligned}\Pr[\text{not cover}] &= \\ \Pr[\exists \text{ element not covered}] &\leq \\ \sum_e \Pr[e \text{ not covered}] &\end{aligned}$$

**For one element e  
and for one iteration**

$$\Pr[e \text{ not covered}] < 1/e$$

**For one element e  
and for all iterations together**

$$\Pr[e \text{ not covered}] < (1/e)^{\ln(n)+3} = \frac{1}{e^3 n}$$

$$\sum_e \Pr[e \text{ not covered}] < n \frac{1}{e^3 n} = \frac{1}{e^3} < 0.05$$

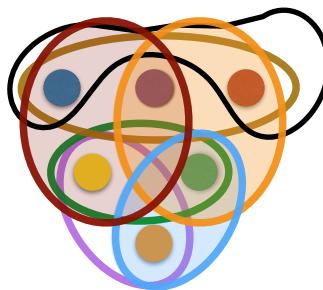
**So:**

$$\Pr[\text{cover}] > 0.95$$

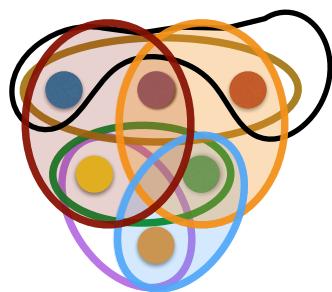
# **Result**

**Iterated randomized rounding gives  
collection of sets  
that is a set cover with  
probability 95% and  
with average cost  
at most  $(\ln(n)+3)$  OPT.**

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



# Result so far

Iterated randomized rounding gives collection of sets

that is a set cover with probability 95%  
and with average cost at most  $(\ln(n)+3)$  OPT.

Not guaranteed!

Not guaranteed!

**Q: What if you want the output  
to always be a set cover?**

**Desired result:**  
algorithm that gives  
collection of sets  
that **is** a set cover  
and with **average**  
cost at most  $O(\ln(n))$  OPT.

**Guaranteed!**

**Obvious solution:**  
Replace  
“repeat  $\ln(n)+3$  times”  
by  
“repeat until you have a set cover.”

# Equivalent algorithms

Repeat  
For each  $S$   
Put  $S$  in cover w.pr.  $x(S)$   
(if not there yet)  
Until you have a set cover

Repeat  
Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$   
Put  $S$  in cover  
(if not there yet)  
Until you have a set cover

# Sample-and-iterate algorithm

**Repeat**

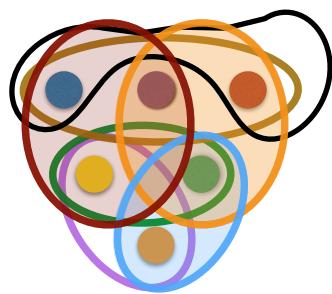
**Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$**

**Put  $S$  in cover**

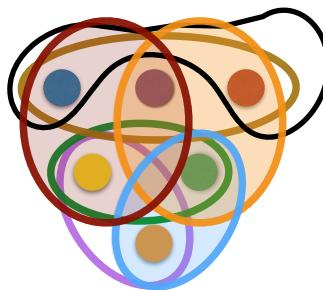
**(if not there yet)**

**Until you have a set cover**

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



# Sample-and-iterate algorithm

**Repeat**

**Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$**

**Put  $S$  in cover**

**(if not there yet)**

**Until you have a set cover**

**Repeat**

**Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$**

**Put  $S$  in cover**

**(if not there yet)**

**Until you have a set cover**

**Analysis**

### **1. Expected cost of output**

$T = \text{\#iterations (stopping time)}$

$C_t = \text{cost of set chosen at iteration } t$

**Cost of output:**  $\sum_{t=1}^T C_t$

**Repeat**

**Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$**

**Put  $S$  in cover**

**(if not there yet)**

**Until you have a set cover**

**Cost of output:**  $\sum_{t=1}^T C_t$

**Expected cost of output:**  $E[\sum_{t=1}^T C_t]$

**NB:** cannot exchange  $E[ ]$  and summation here!

# **Wald's equation**

**T stopping time**

**X<sub>t</sub> random variable**

**If X<sub>t</sub> bounded from above:**

$$\text{then } E\left[\sum_{t \leq T} X_t\right] \leq \mu E[T]$$

**NB: can now exchange E[] and summation!**

**Expected cost of set chosen at iteration t:**

$$E[C_t] = \sum_S \Pr[S \text{ chosen}] c_S = \frac{\sum_S c_S x_S}{\sum_S x_S} = \mu$$

**Expected cost of output:**  $E[T]\mu = E[T] \frac{\sum_S c_S x_S}{\sum_S x_S}$

**Next problem: What is  $E[T]$ ?**

Repeat

Choose  $S$  w.pr.  $x_S / \sum_{S'} x(S')$

Put  $S$  in cover

(if not there yet)

Until you have a set cover

More  
notations

## 2. Expected number of iterations

$n_t = \#\text{elts not yet covered after } t \text{ iterations}$

$$n_0 = n, n_T = 0, n_{T-1} \geq 1$$

# Wald's equation for dependent decrements

Consider a random decreasing sequence

$$n_0, n_1, \dots, n_T,$$

where  $T$  is a stopping time. If

$$E[n_t - n_{t+1} | n_t] \geq f(n_t),$$

where  $f$  is an increasing function, then

$$E[T] \leq 1 + E\left[\int_{n_{T-1}}^{n_0} \frac{1}{f(z)} dz\right].$$

To bound  $E[T]$ , analyze  
#elts not yet covered after  $t$  iterations

**Analyze**  $n_t$

**Analyze**  $E[n_t - n_{t+1} | n_t]$ : elts covered in next iteration

**Fix an element**  $e$  among the  $n_t$

$\Pr[e \text{ covered in next iteration}] =$

$\Pr[S \text{ chosen contains } e] =$

$$\frac{\sum_{S:e \in S} x_S}{\sum_{S'} x_{S'}} \geq \\ 1 / \sum_{S'} x_{S'}$$

**Sum over**  $e$

$$E[n_t - n_{t+1} | n_t] \geq f(n_t)$$

$$E[n_t - n_{t+1} | n_t] \geq n_t / \sum_{S'} x_{S'}$$

## Wald's equation for dependent decrements

Consider a random decreasing sequence

$$n_0, n_1, \dots, n_T,$$

where T is a stopping time. If

$$E[n_t - n_{t+1} | n_t] \geq f(n_t),$$

where f is an increasing function, then

$$E[T] \leq 1 + E\left[\int_{n_{T-1}}^{n_0} \frac{1}{f(z)} dz\right].$$

## Application:

Stopping time  $E[T] \leq 1 + \int_1^n \frac{\sum_S x_S}{z} dz = 1 + \sum_S x_S \ln(n)$

# Together

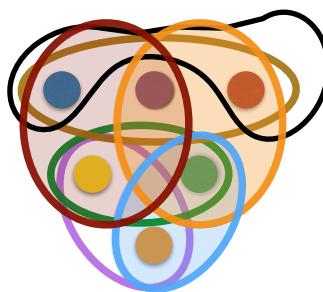
## Expected cost of output:

$$(1 + (\sum_S x_S) \ln(n)) \frac{\sum_S c_S x_S}{\sum_S x_S} \leq (1 + \ln(n)) \text{ OPT}$$

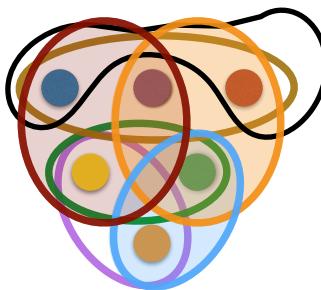
### Result:

the algorithm gives  
a collection of sets that is a set cover  
with average cost at most  $(1 + \ln(n)) \text{ OPT}$ .

# Set cover, linear programming and randomized rounding



# Set cover, linear programming and randomized rounding



# Result

The sample-and-iterate algorithm gives a collection of sets that is a set cover with average cost at most  $(1 + \ln(n)) \text{ OPT}$ .

# **A more efficient algorithm**

**Linear programming takes  
polynomial time  
but  
is often slower than  
combinatorial algorithms**

# **Greedy**

**Repeat**

**Choose  $S$  maximizing  $\#(\text{new elts covered})/c_S$**

**Put  $S$  in cover**

**Until you have a set cover**

**Result:**  
Greedy also gives  
a collection of sets  
that **is** a set cover  
and with  
**cost at most  $(1 + \ln(n)) \text{ OPT.}$**

**Can we do better?**

**No:**

**It is NP-hard to obtain (in polynomial time) a  
better-than- $\ln(n)$  approximation  
for set cover**



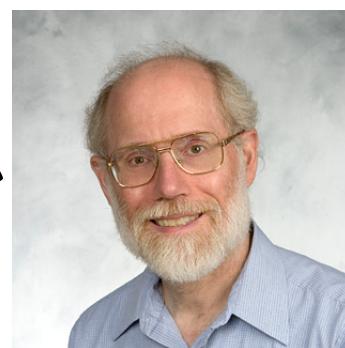
**Uri  
Feige**



**Vašek  
Chvátal**



**László  
Lovász**



**David  
Johnson**



**Neal  
Young**

# What have we learned?

- Famous problem: set cover
- Concept: Randomization
- Algorithmic technique: Randomized rounding
- Analysis tool: Linearity of expectation
- Laying out an analysis: slow & steady, orderly - like hiking up a mountain

# Set cover, linear programming and randomized rounding

