

Mining Concepts from Code

using Community Detection in co-occurrence Graphs

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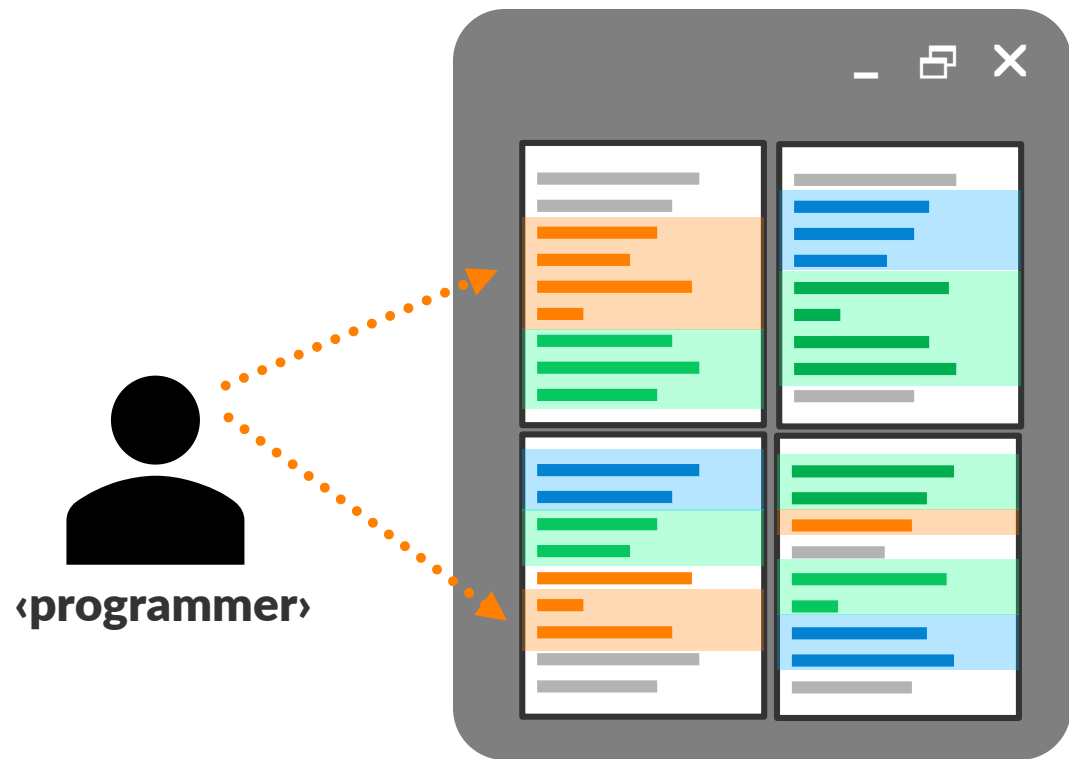
Problem: Architectural Drift

With growing code bases...

- » Concepts tend to **scatter** and **entangle**
- » Programmers need **more time** to locate concepts



Goal: Concept Recovery and Location



Name-based Concept Model

concept locations

which concept a name belongs to

Canvas » draw: anObject

^ anObject drawOn: self

Morph » drawOn: aCanvas

aCanvas fillRectangle: self bounds.

Morph » bounds: newBounds

self position: newBounds topLeft;

extent: newBounds extent.

concepts

prevalent names

draw, canvas, fill, ...

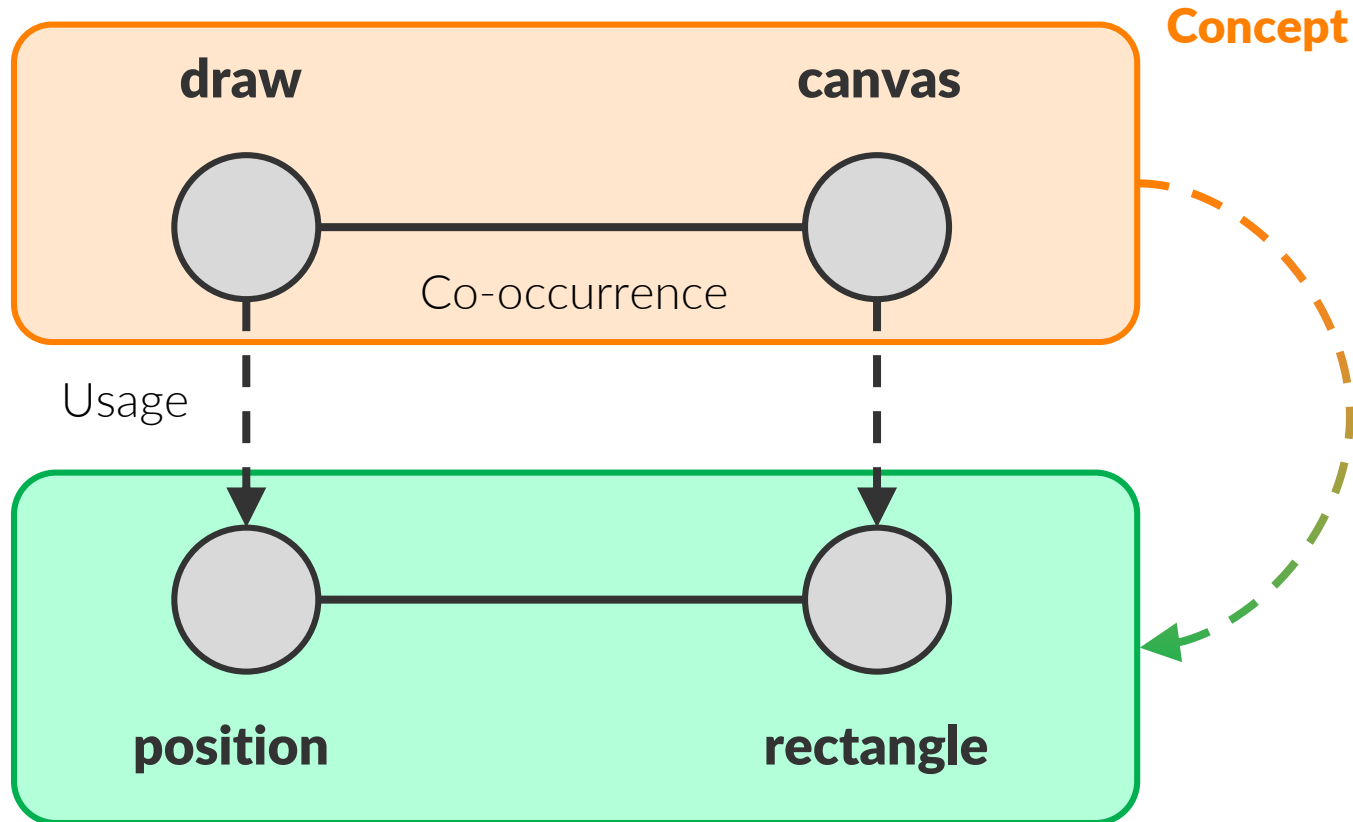
relations

(e.g. usage)

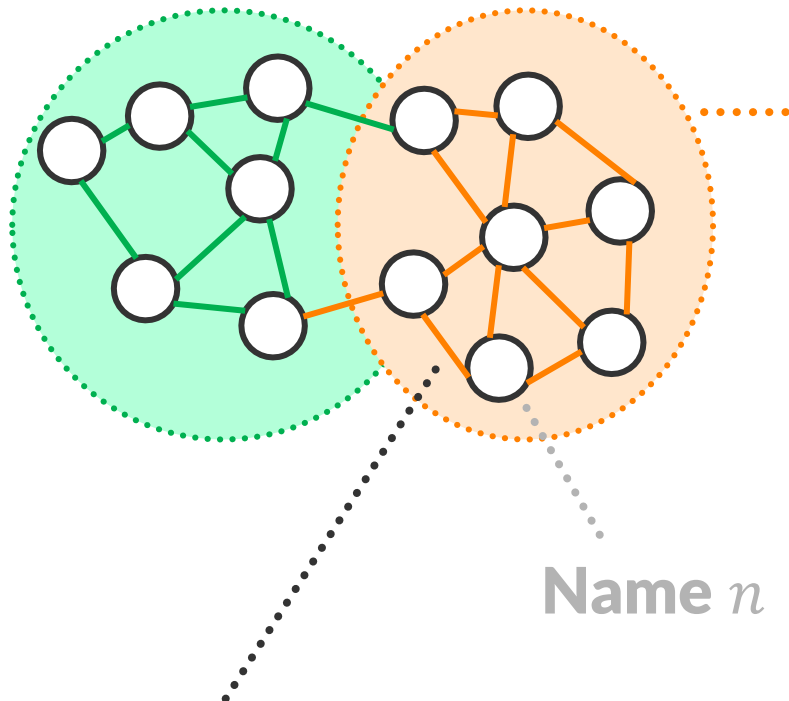
bounds, position,
extent, ...

Graph-based Semantic Models

Nodes are **names**. **Edges** indicate they **co-occur** in close proximity.



Random Graph Model + Topic Model

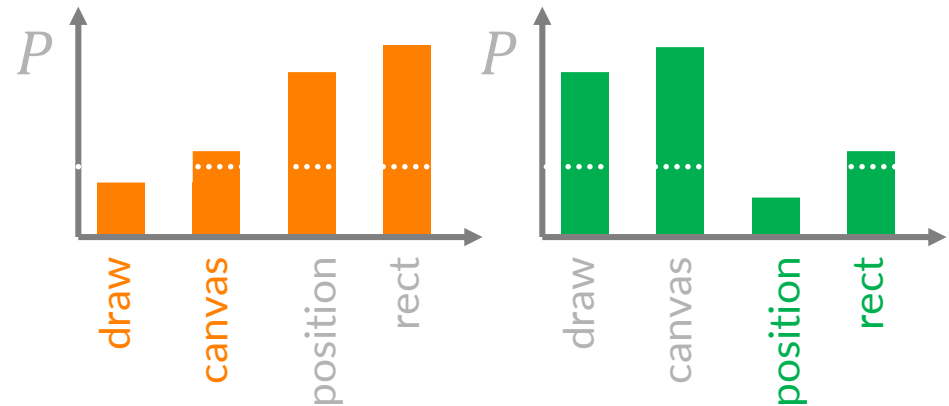


Concept c

distribution over names $P(n|c)$

global frequency $P(c)$

Name n

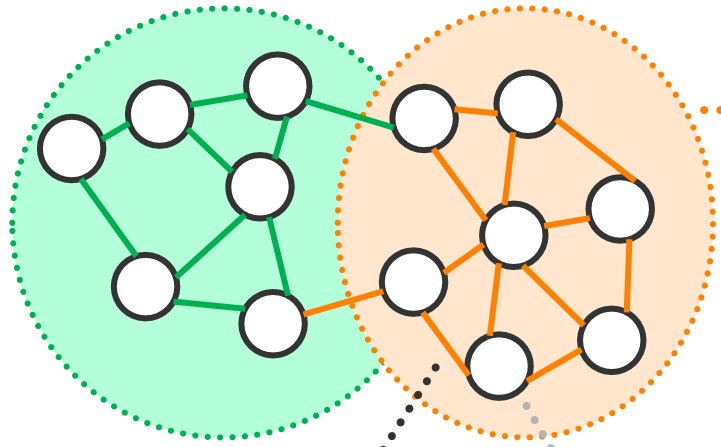


Edge e

$$P(e = (n_1, n_2) | c) \propto P(n_1|c)P(n_2|c)$$

$$P(G = (V, E) | \dots) \propto \prod_{e \in E} \sum_c P(e|c) =$$

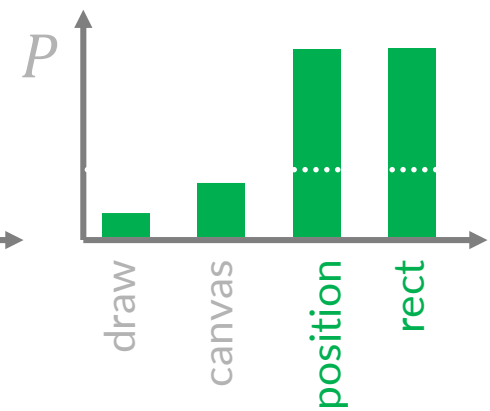
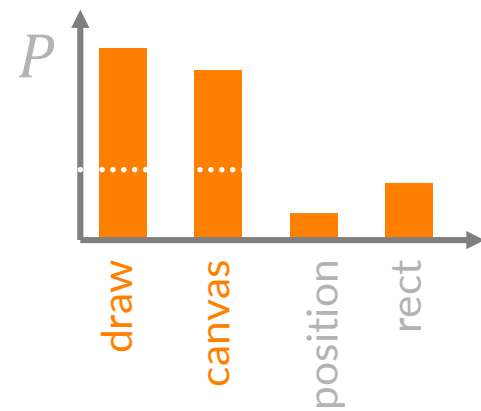
Random Graph Model + Topic Model



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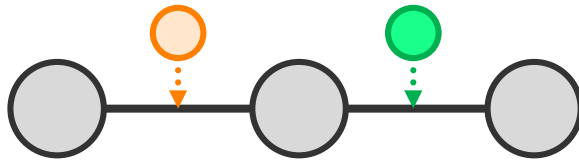
Edge e

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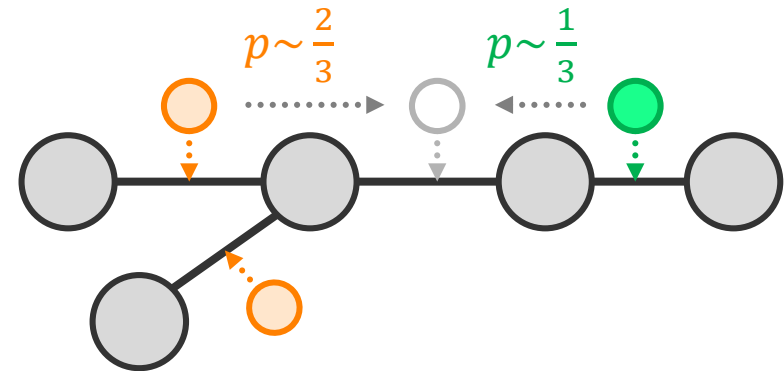
$$P(G = (V, E) | \dots) \propto \prod_{e \in E} \sum_c P(e|c) =$$

Determining $P(n|c)$ via Gibbs Sampling

Random assignment
of latent variables c to edges



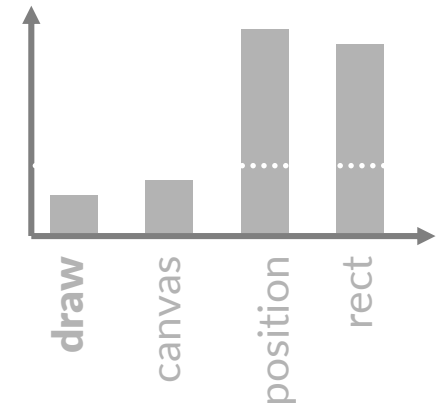
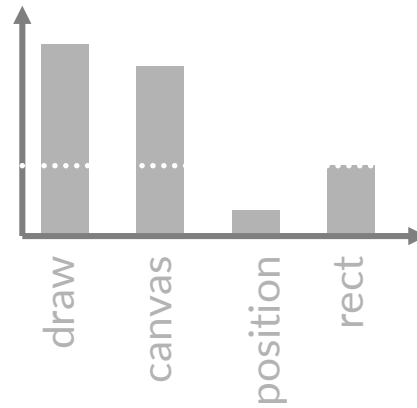
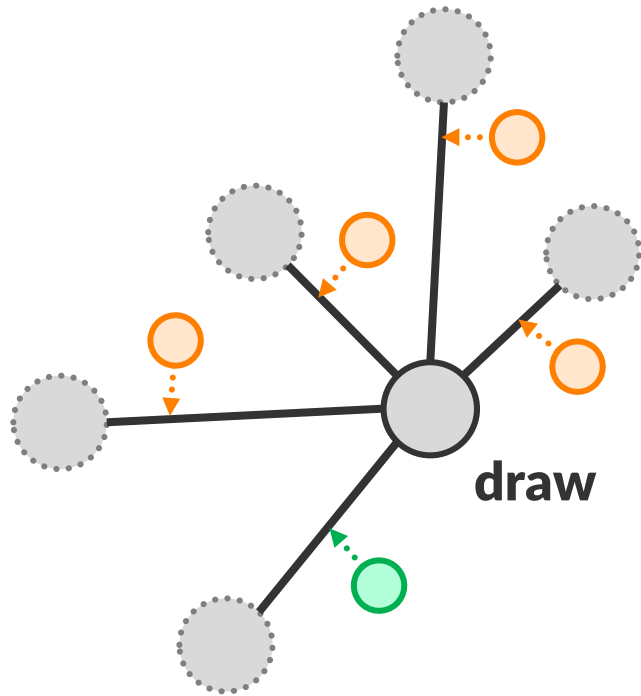
Iterative Re-sampling



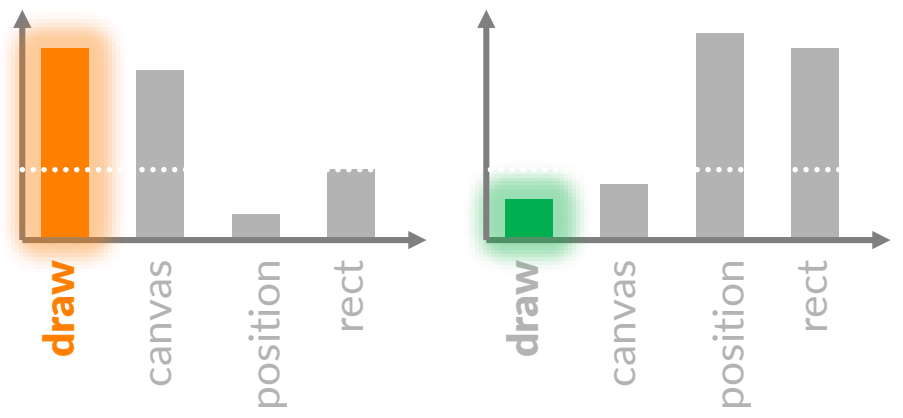
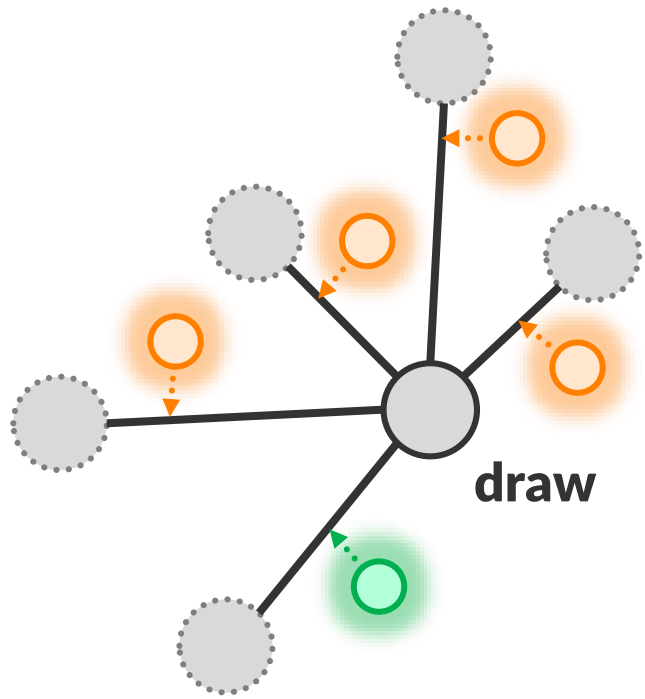
1. Decide on maximum number of concepts
2. Uniformly assign a concept to each edge
3. Re-assign each edge until near convergence

(clustering edges instead of nodes)

Random Graph Model + Topic Model

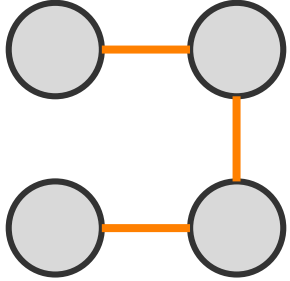


Random Graph Model + Topic Model

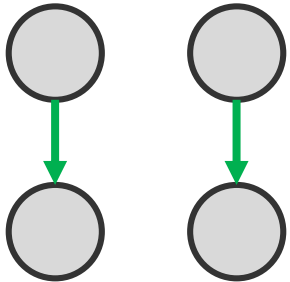


Multi-view Concepts

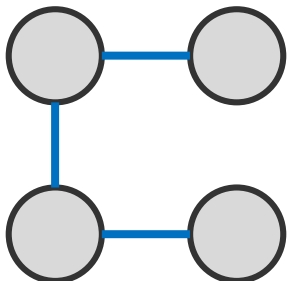
Co-located Names



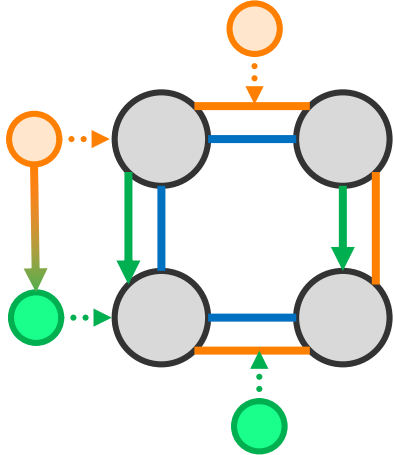
Run-time Call Data



Git Commit (Diff)

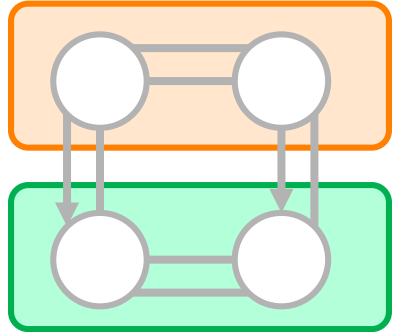


(Multi-)Graph



Concept Labeling

Concept Distribution



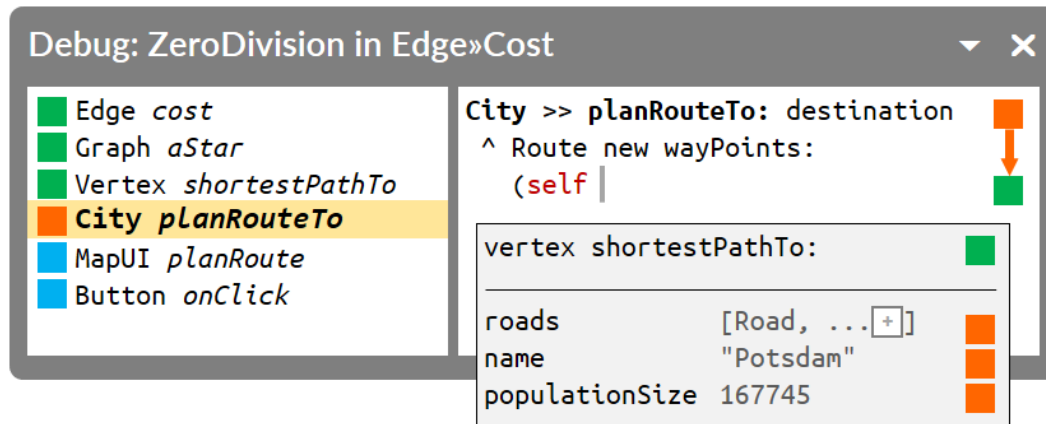
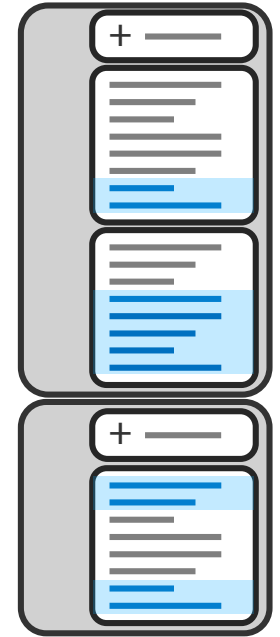
Future Work: Concept-aware Tooling

- » **Highlight** concepts
- » **Improve relevance** of information displayed during
 - › search
 - › code completion
 - › debugging

graph, vertex, node

city, road, speed

draw, canvas, fill, ...

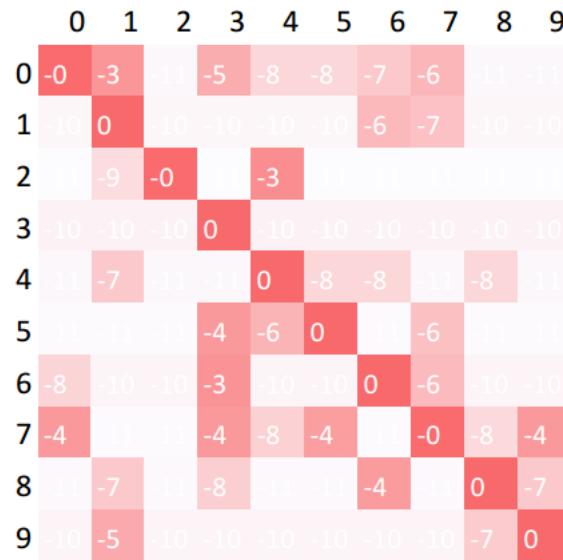


Concept Coherence (Mimno et al.)

Project	concepts	LDA				Co-occurrence Graph			
		C_2	C_4	C_8	C_{12}	C_2	C_4	C_8	C_{12}
EPIC	10	-1.5	-11.0	-53	-135	-1.0	-7.0	-43	-114
	15	-1.6	-9.9	-53	-143	-0.9	-8.0	-46	-123
	20	-1.7	-10.6	-57	-144	-1.2	-8.1	-47	-125
	25	-1.8	-11.7	-56	-144	-1.2	-9.0	-48	-126
Django	10	-1.9	-12.1	-65	-166	-1.5	-10.0	-51	-135
	15	-2.4	-13.4	-68	-171	-1.3	-9.8	-56	-143
	20	-2.3	-12.6	-67	-170	-1.3	-10.7	-57	-144
	25	-2.0	-12.3	-69	-173	-1.4	-10.3	-58	-148
IPython	10	-2.0	-13.9	-67	-172	-1.9	-12.1	-60	-145
	15	-1.9	-12.8	-68	-167	-1.4	-10.3	-57	-150
	20	-2.1	-13.1	-70	-169	-1.4	-9.2	-55	-144
	25	-1.7	-12.2	-63	-164	-1.5	-10.4	-56	-142

■ **Table 4** Concepts inferred from the *EPIC* digital simulator

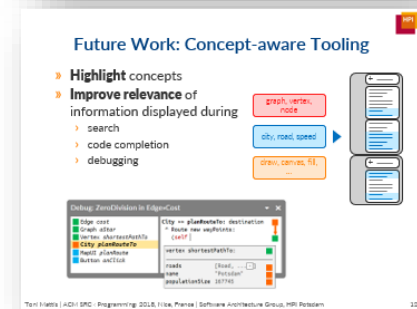
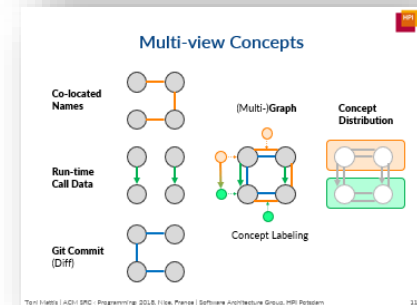
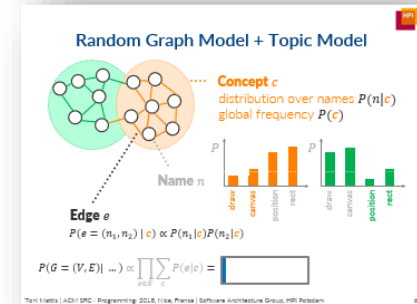
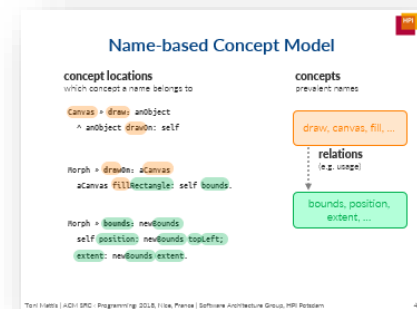
	Names	Comment (Interpretation)
0	<i>event if true mouse hand</i>	A cross-cutting concept handling mouse interaction
1	<i>morph layout panel create box</i>	The editor window
2	<i>canvas draw box center color</i>	Drawing circuitry
3	<i>is simulation if event not</i>	Event-driven simulator
4	<i>color rectangle string fill at</i>	Drawing shapes and text
5	<i>wire point anchor points bundle</i>	Wires, bundles of wires, and their connections
6	<i>input output values first with</i>	Expectations encoded in unit tests
7	<i>components panel component command all</i>	The panel containing pre-defined components
8	<i>xml circuit element named as</i>	The (de)serializer
9	<i>file name stream as named</i>	File reading/writing



■ **Figure 4** Abstract concepts (left) and how likely they relate to implementation-specific concepts (top). Values are logarithmically scaled.

Summary

- » Graph-based concept modeling is a **framework** based on a co-occurrence relation over names
- » **Future work:** extend tools to exploit conceptual information
- » By giving programmers **feedback** how well their modules align with concepts, they can **counteract** architectural drift



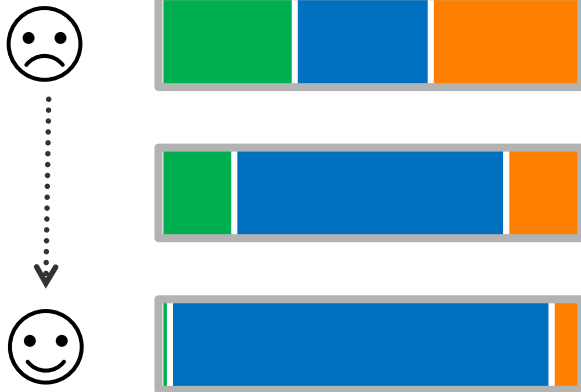
Backup Slides

A Perspective on Modularity

module entropy:

tangling

module



$$H(m) = - \sum_c p(c|m) \log_2 p(c|m)$$

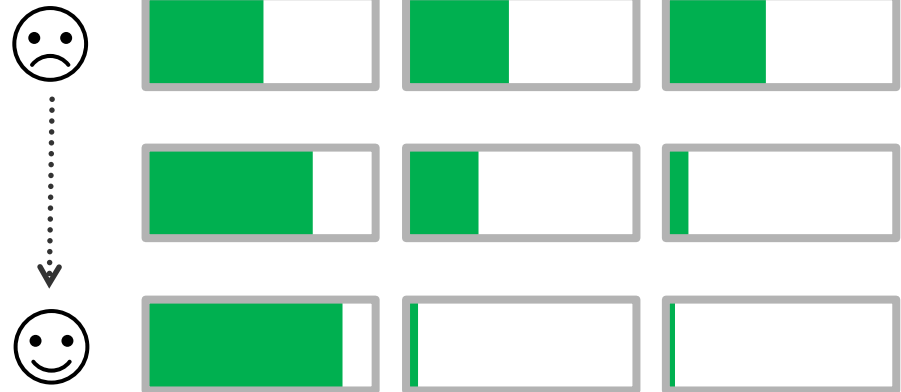
concept entropy:

scattering

module

module

module



$$H(c) = - \sum_m p(m|c) \log_2 p(m|c)$$

...high values indicate need for refactoring or cross-cutting concerns

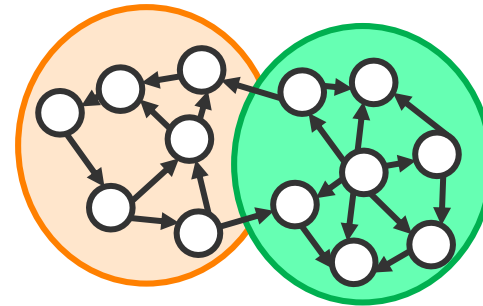
E. Linstead, P. Rigor, S. Bajracharya, C. Lopes, and P. Baldi, "Mining Concepts from Code with Probabilistic Topic Models," ASE, 2007

Related Work

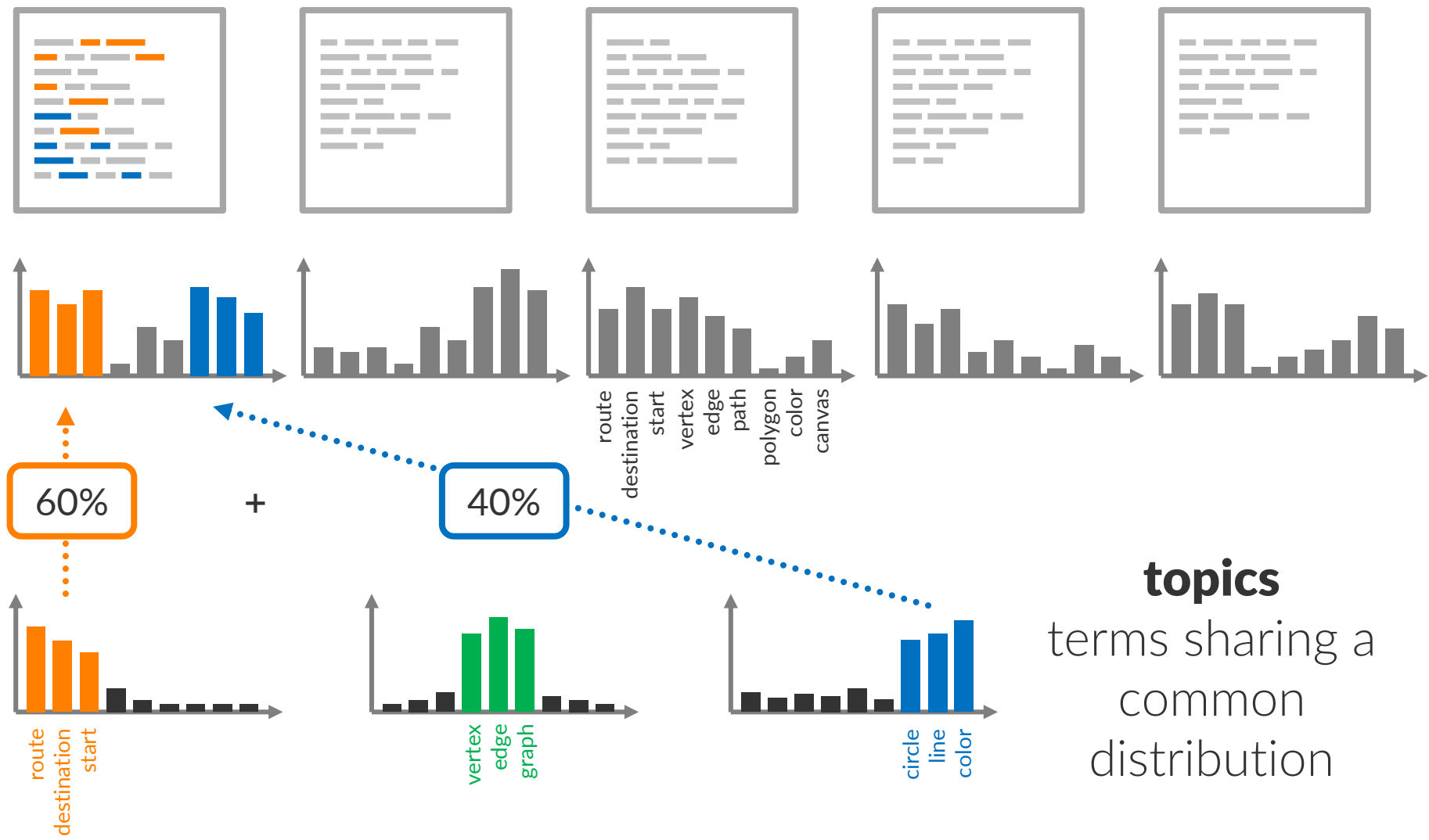
Topic Models



Random Graph Models with Community Structure



Topic Models



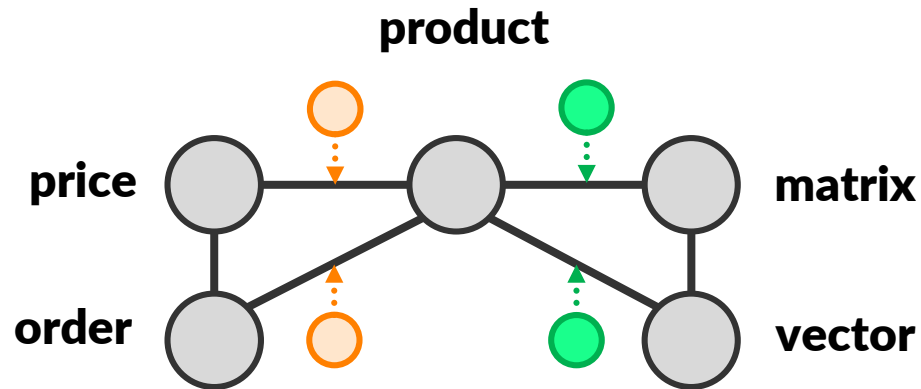
topics
terms sharing a
common
distribution

Disambiguating Names

« product »

```
order.total += product.price;
```

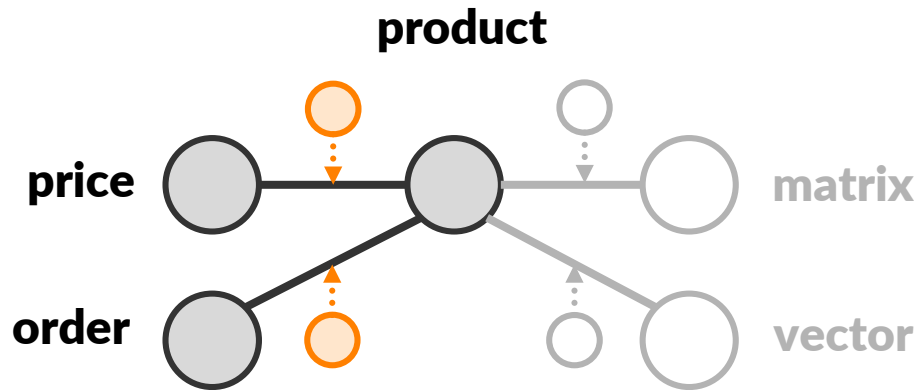
```
product = matrix * vector;
```



Disambiguating Names

`order.total += product.price;`

`product = matrix * vector;`



Disambiguating Names

`order.total += product.price;`

`product = matrix * vector;`

