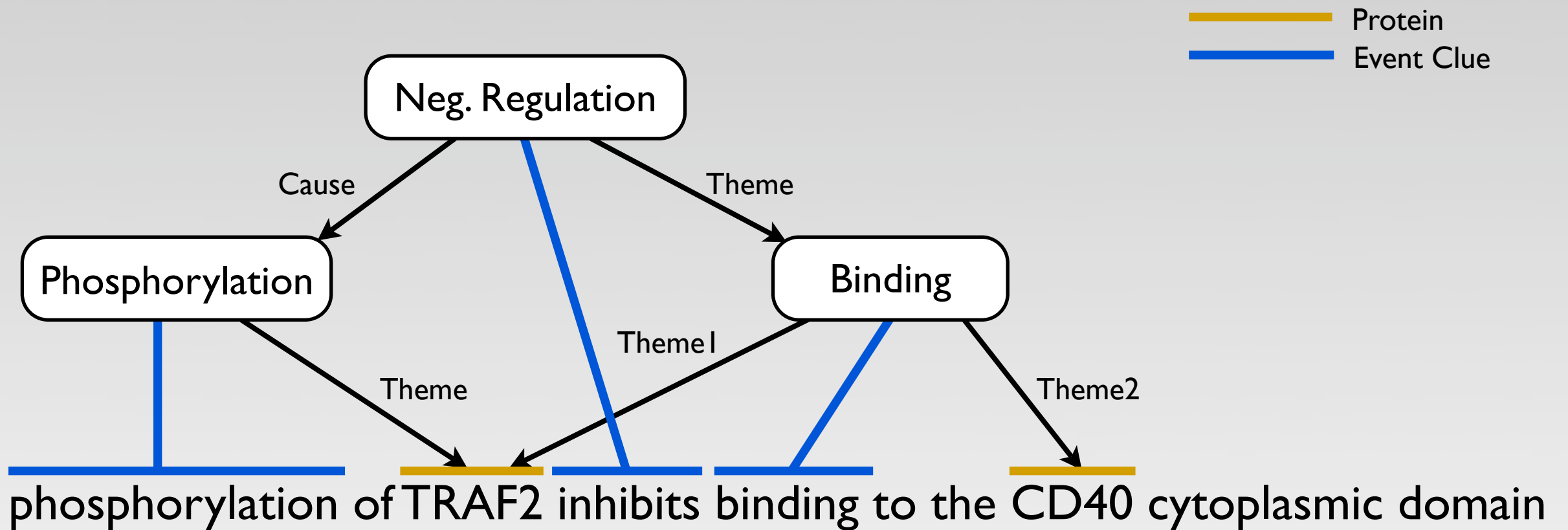


# A Markov Logic Approach to Bio-Molecular Event Extraction

Sebastian Riedel, Hong-Woo Chun, Toshihisa Takagi, Jun`ichi Tsujii

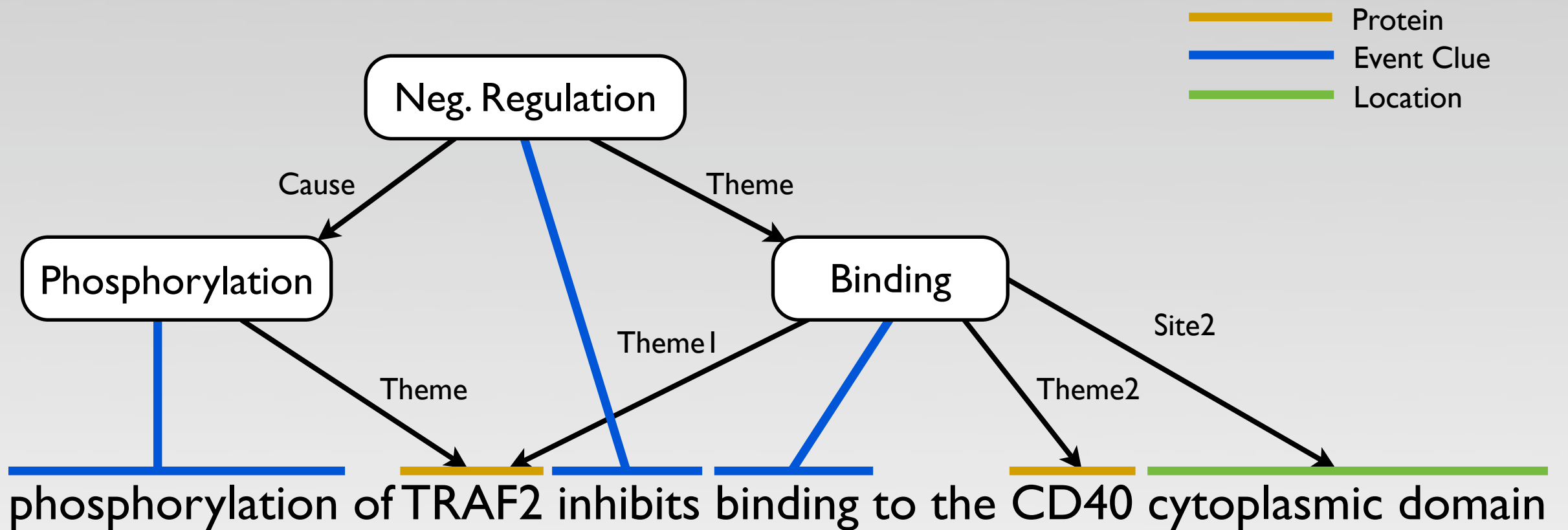


# Task I



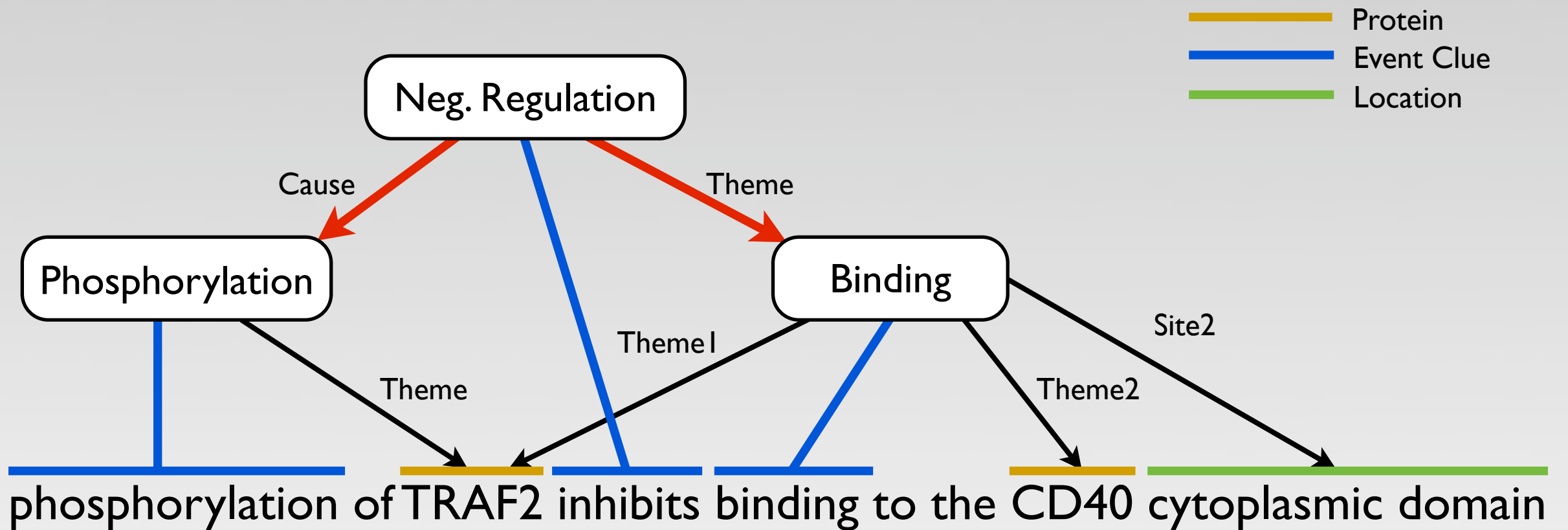
- Find event **clues** and **arguments**

# Task 2



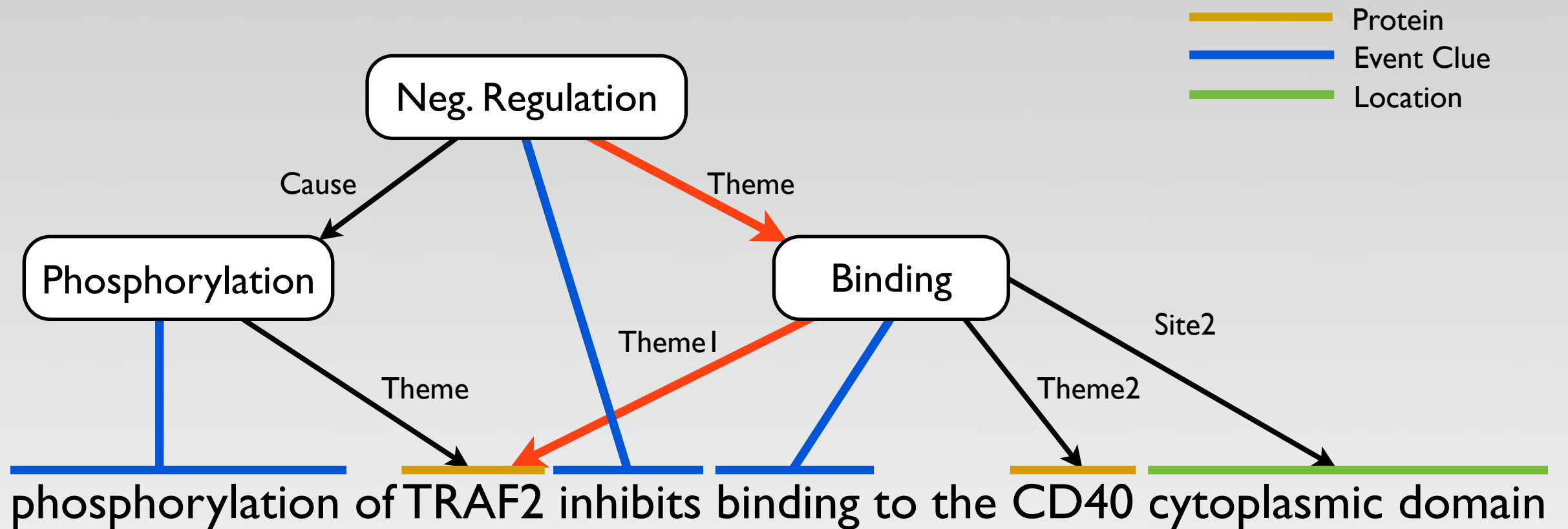
- Find and attach **cellular locations**

# Observations



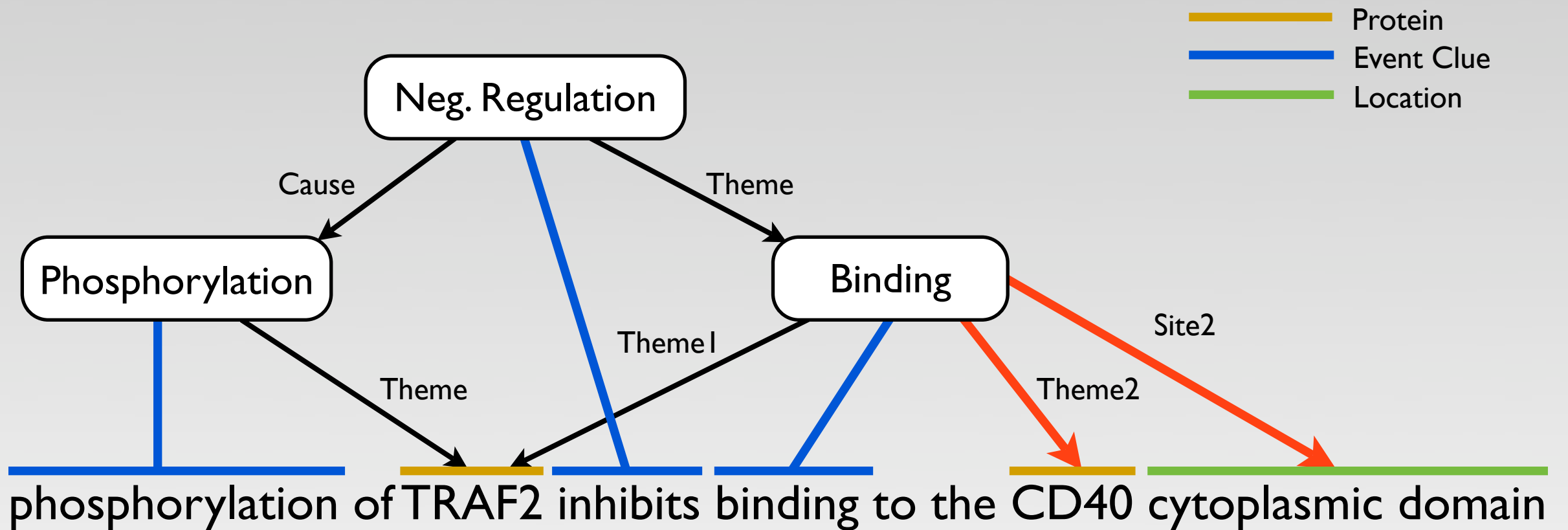
- At least one **Theme**

# Observations



- Regulation **transitively** involves proteins

# Observations



- **Site2** arguments require **Theme2**

# Approach

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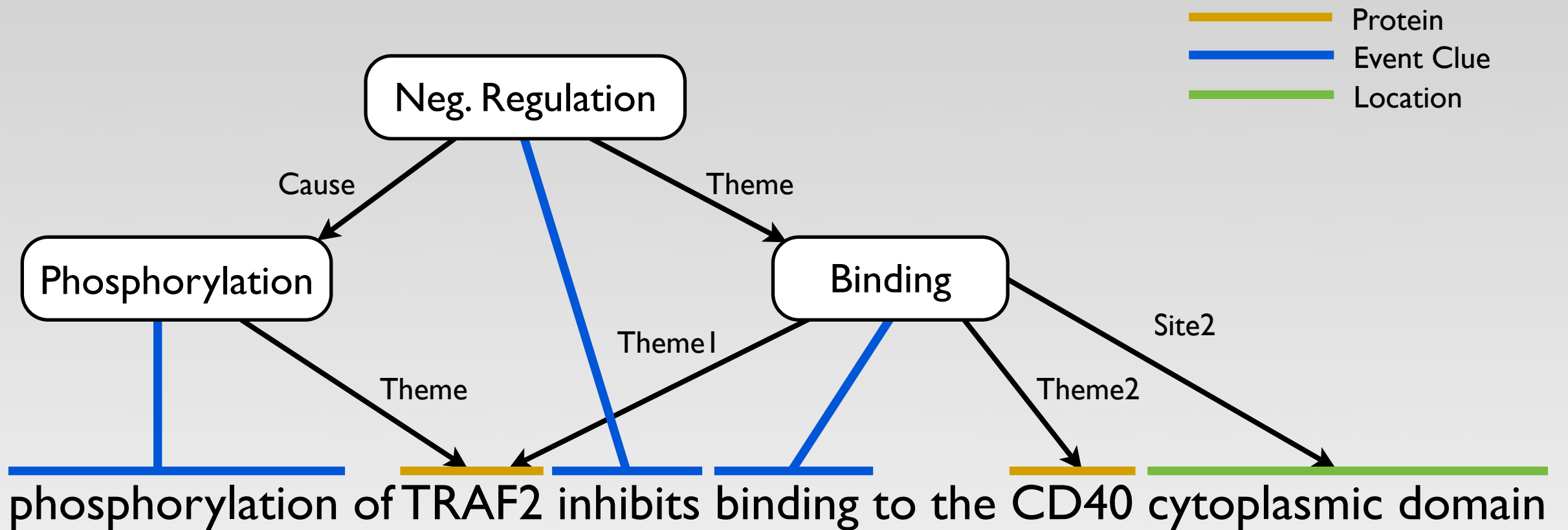
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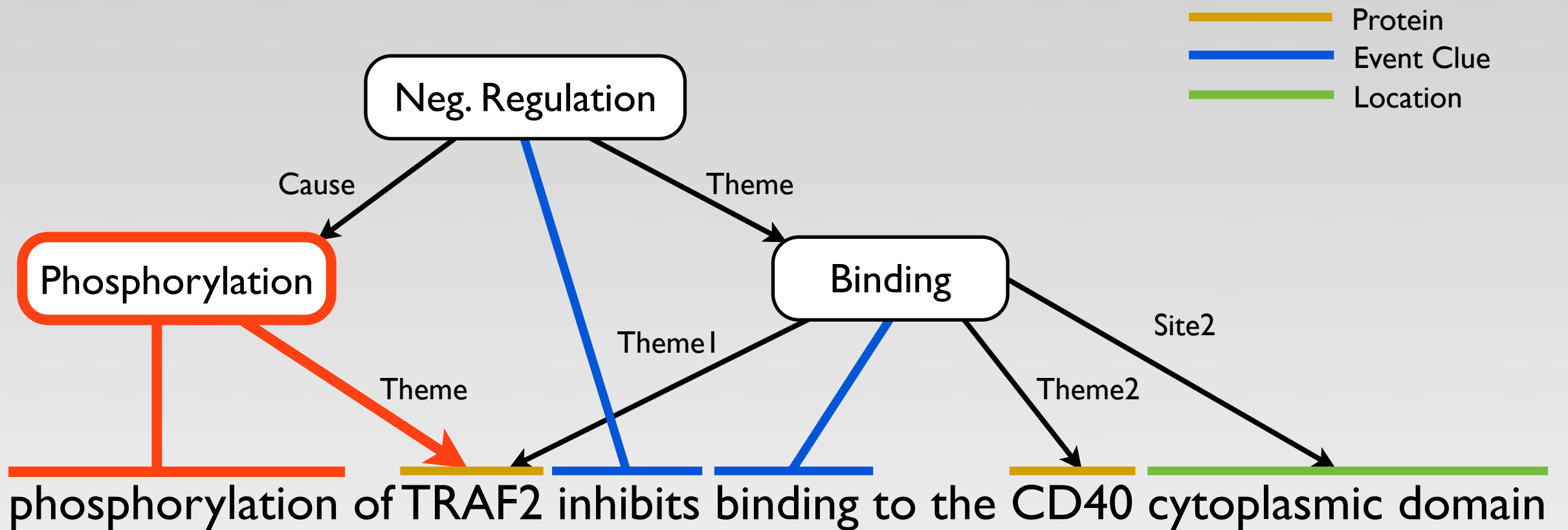
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  - Map to **link structure** over tokens

# Event Prediction



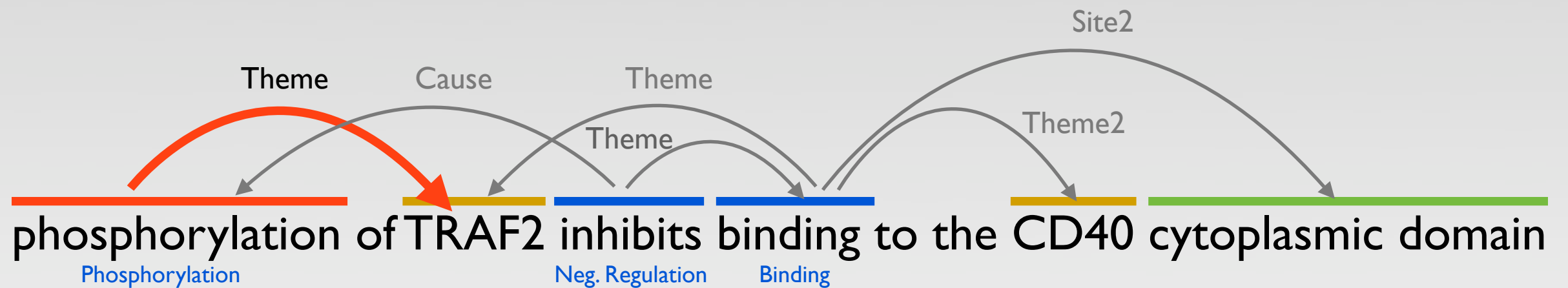
- Event structure where events are entities

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# Link Prediction



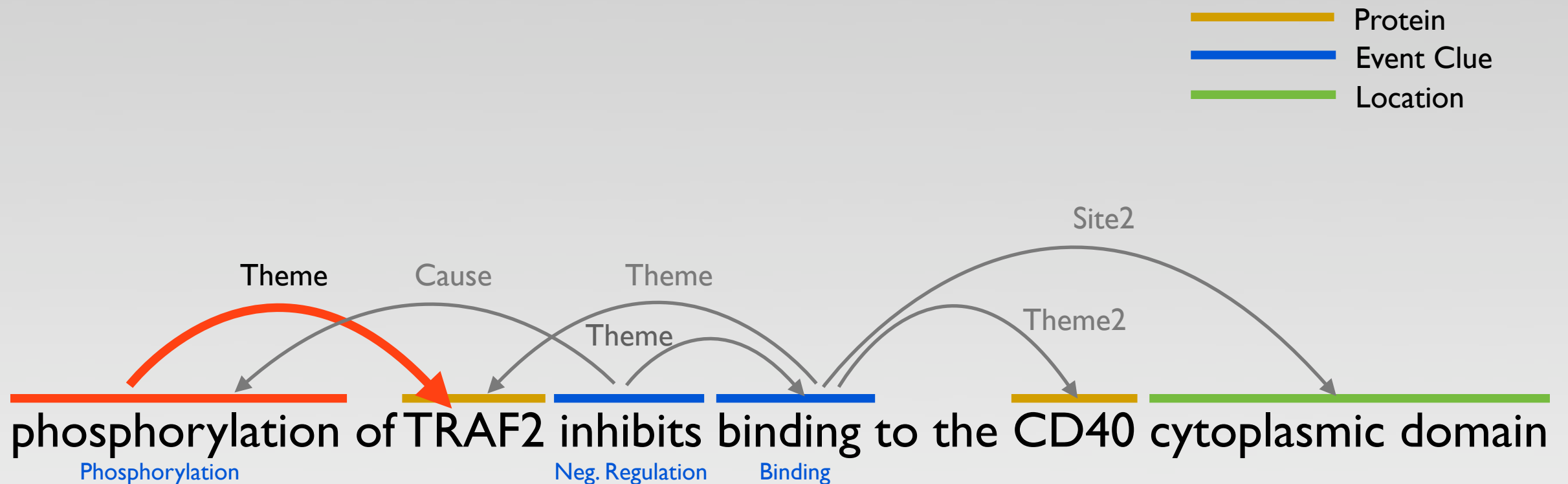


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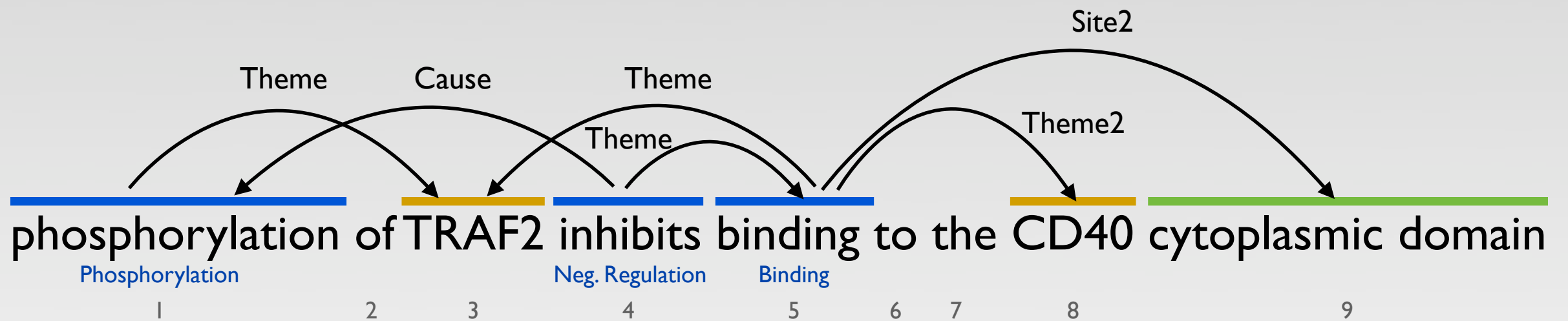
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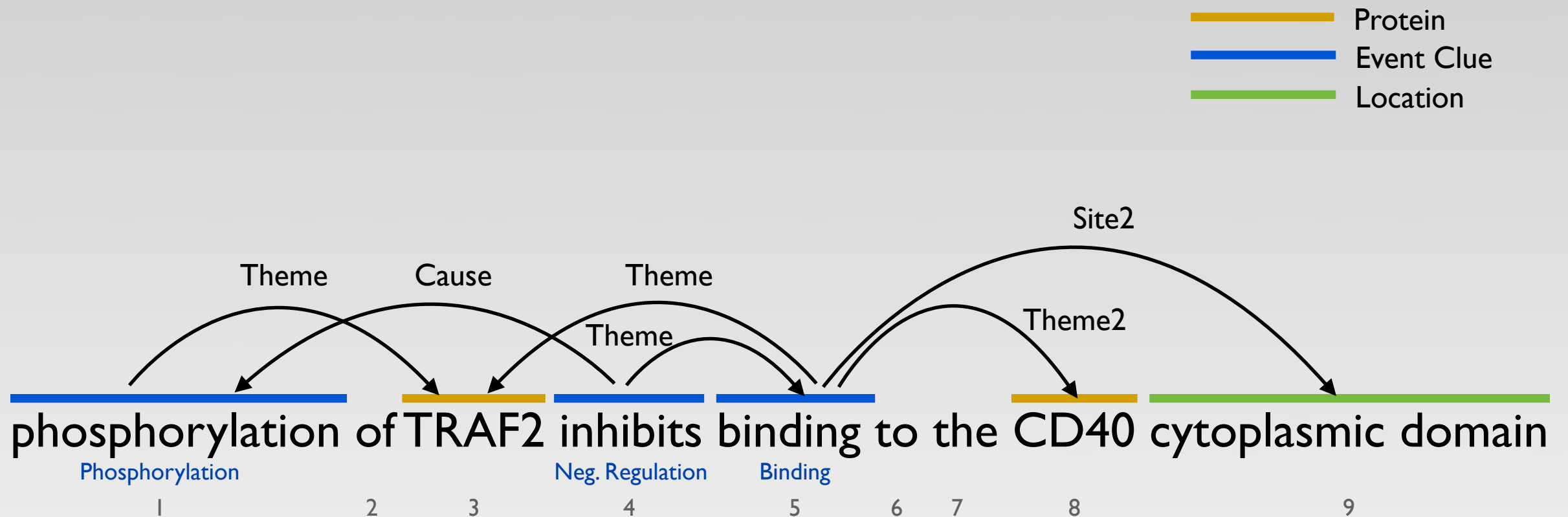


- Events are projected on their clues
- Close to **Semantic Role Labelling** : cf Meza-Ruiz & Riedel, 2009)

# Markov Logic

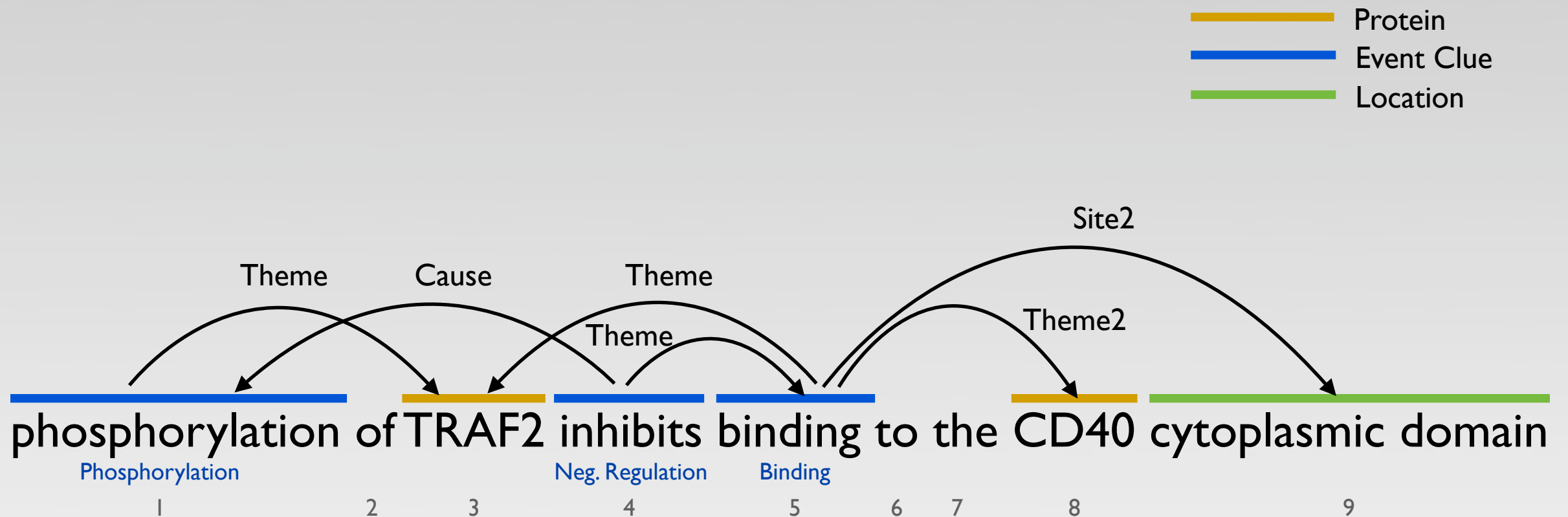


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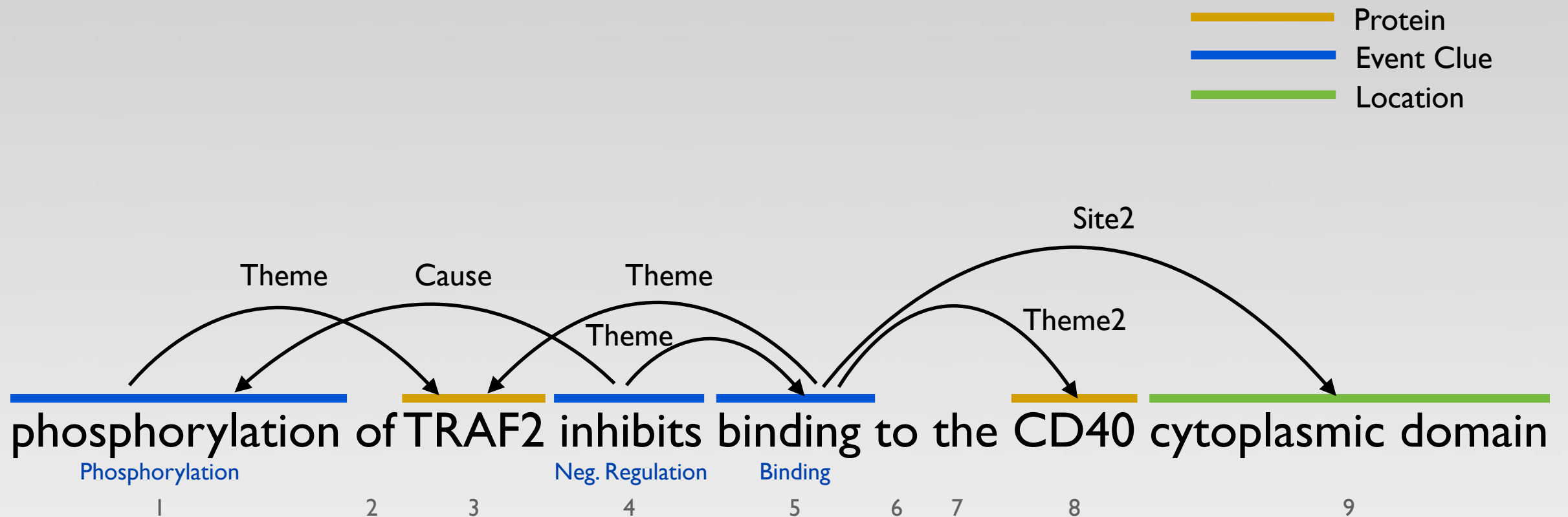
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{*event*(1,Phosph.), *role*(1,3,Theme), *site*(9)...}

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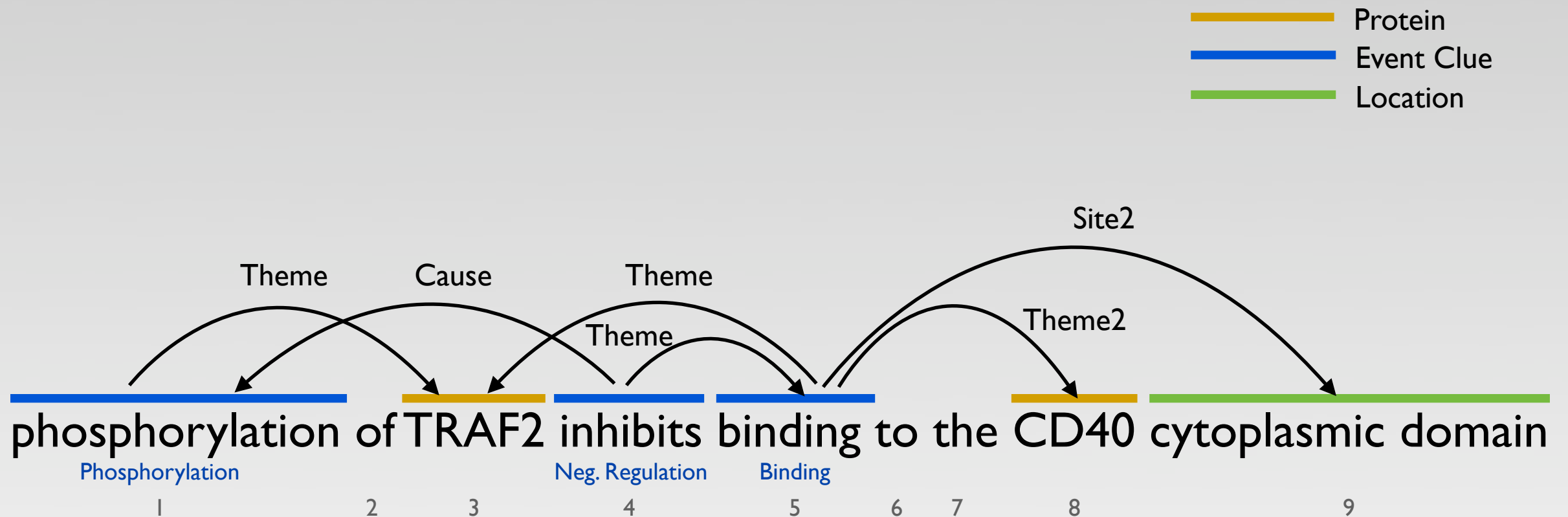
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# Clue Detection

- Protein
- Event Clue
- Location

phosphorylation of TRAF2 inhibits binding to the CD40 cytoplasmic domain

1                    2                    3                    4                    5                    6                    7                    8                    9





# Clue Detection



phosphorylation of TRAF2 inhibits binding to the CD40 cytoplasmic domain

1                    2                    3                    4                    5                    6                    7                    8                    9

*(Note: A red oval highlights the word "inhibits" and the blue line above it, with "Neg. Regulation" written below it.)*

- Distinguish between good and bad worlds  
 $word(i, inhibits) \Rightarrow event(i, Neg.Regulation) <1.2>$

# Site Extraction

— Protein  
— Event Clue  
— Location

phosphorylation of TRAF2 inhibits binding to the CD40 cytoplasmic domain

1 2 3 4 5 6 7 8 9

- **Copy and paste** formulae:  
 $word(i,w) \Rightarrow location(i) \langle weight(w) \rangle$

# Compact Representation



phosphorylation of TRAF2 inhibits binding to the CD40 cytoplasmic domain

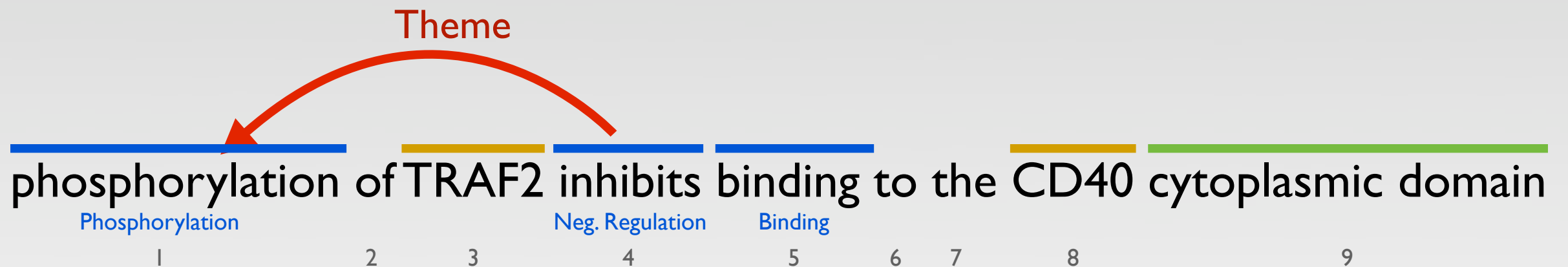
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*(Note: In the original image, the word "inhibits" is circled in red, and a blue line above it is labeled "Neg. Regulation".)*

- Compact representation:

$word(i,w) \Rightarrow event(i,t) \langle weight(w,t) \rangle$

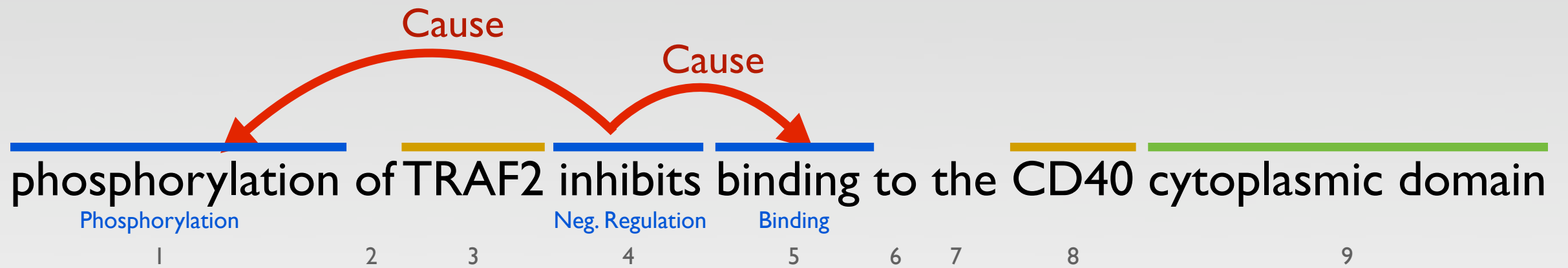
# Argument Extraction



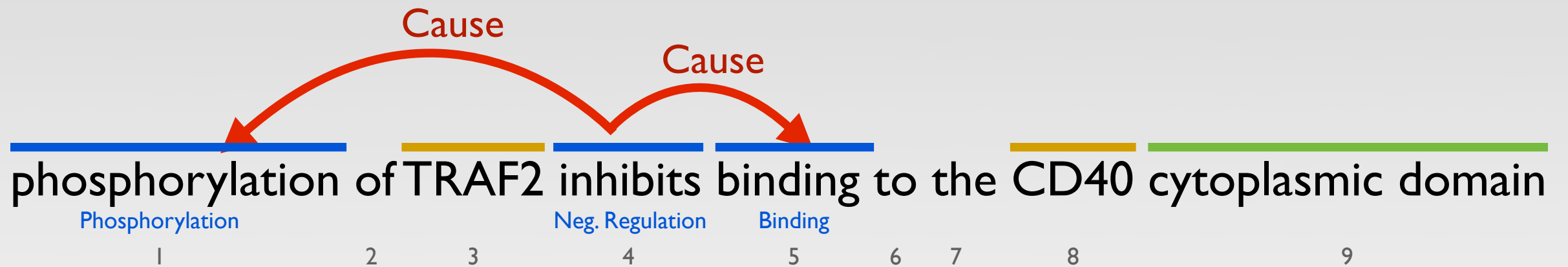
- Subjects are themes:

$$dep(i,j,subj) \Rightarrow role(i,j,Theme) <1.23>$$

# Global Correlations



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- There need to be Themes:

$$event(e,t) \Rightarrow \exists a. role(e,a,Theme) \langle 999.9 \rangle$$

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possible world      formula      weight      binding of free variables

ground feature

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- Fast to train: **3 hours** on MacBook Pro

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# Impact of Joint Inference (on Atom FI)

Predicate	CORE	VALID	FULL
<i>total</i>	50.7	60.1	61.9
<i>event</i>	52.8	63.2	64.3
<i>role</i>	44.0	53.5	55.7
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- FULL does not explicitly consider *site*
- But Turku still did better w/ a local model

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