

# **KNOW2LOOK: COMMONSENSE KNOWLEDGE FOR VISUAL SEARCH**

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# IMAGE RETRIEVAL

User Query

Concrete

Q: *bicycle in street*

Abstract

Q: *environment friendly traffic*

# IMAGE RETRIEVAL

User Query

Concrete



*Q: bicycle in street*

Abstract

*Q: environment friendly traffic*

*"Wow! Double-decker buses still run!"*

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Q: *bicycle in street*

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Text-only



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Q: *bicycle in street*

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“Wow! Double-decker buses still run!”

Visual objects: bicycle, bus, car

Text-only



Text + visual



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“Wow! Double-decker buses still run!”

Visual objects: bicycle, bus, car

“Biking by the river”

Visual objects: train, piano

Text-only 

Text + visual 

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Text-only X

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Text + visual ✓

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Visual objects: bicycle, bus, car



*"Biking by the river"*

Visual objects: train, piano



*"Riding for a cause."*

Visual objects: person, bicycle

Text-only

Text-only

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*"Riding for a cause."*

Visual objects: person, bicycle  
CSK: (riding bicycle, be, environment friendly)

Text/visual

Text + visual + CSK

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Our contribution

# OUTLINE

- CSK: Where do we get it from?
- CSK: How do we use it?
- CSK: How to combine noisy signals?
- CSK: Does it help?

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- Pruned by Jaccard Similarity  $JaccardDistance = 1 - WeightedJaccardSimilarity$

where,

$WeightedJaccardSimilarity =$

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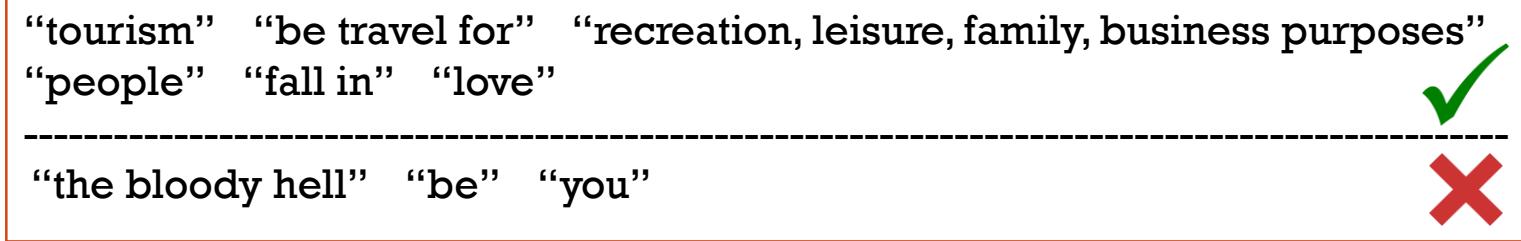
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Domain-specific  
ReVerb triples

- ~22,000 CSK triples

# CSK: HOW DO WE USE IT? FOR QUERY EXPANSION...

- Query string: *travel with backpack*

- CSK to expand query

- t1: (tourists, use, travel maps)
- t2: (tourists, carry, backpack)
- t3: (backpack, is a type of, bag)

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- Document x with features
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  - Visual: person, bag, bottle, bus

Text-only systems 

Text + visual + CSK systems 

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Text + visual + CSK systems



- ✓ CSK bridge vocabulary gap between query and document
- ✓ CSK establish relations between concepts
- ✓ CSK diminish noise from modalities – ensemble effect

# CSK: HOW DO WE USE IT? EXAMPLE...



A tour group is standing on the grass with ruins in the background.  
Group of people standing in front of a stone structure.

Document x



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Textual features  $x_x$



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Query expansion: (an excursion, be trip by, a group of people)

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person, **casual agent**, organism

# CSK: HOW TO COMBINE NOISY SIGNALS? RETRIEVAL RANKING MODEL...

- Mixture LM:

$$P[q|x] = \beta_{CS} P_{CS}[q|x] + (1 - \beta_{CS}) P_{smoothed}[q|x]$$

- Commonsense-aware LM:

$$P_{CS}[q|x] = \prod_i \left[ \frac{\sum_k P[q_i|y_k] P[y_k|x]}{|k|} \right]$$

- Smoothed LM:

$$P_{smoothed}[q|x] = \alpha P_{basic}[q|x] + (1 - \alpha) P[q|B], \text{ where}$$

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CSK triple

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Probabilities based on  
word-wise overlaps

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Background corpus –  
Co-occurring Flickr tags

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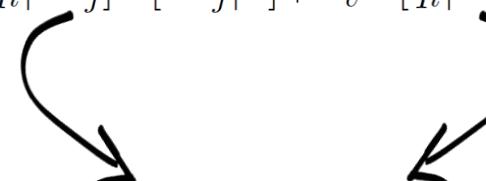
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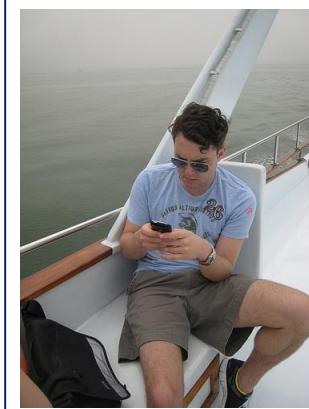
Textual and visual features

# CSK: DOES IT HELP? EVALUATION...

- Image Dataset
  - Flickr30k
  - MS COCO captioned dataset
  - Pascal Sentence Dataset
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Boat trip to see the mythical pink dolphins...  
this is John checking in with the office for that day.

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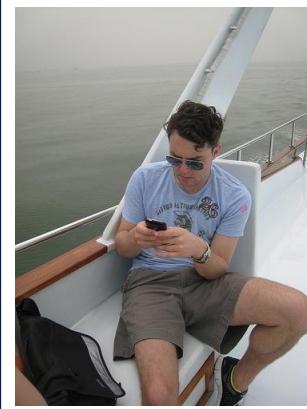
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A group of tourists is crossing a bridge that connects a walking path to a trail of nature.  
Many people cross a very tall footbridge with a tree-covered hill in the background.  
This shows a group of people walking over an arched red bridge.  
People cross a large bridge to get over the body of water.  
People walking over a white and red bridge over a pond.

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social media post



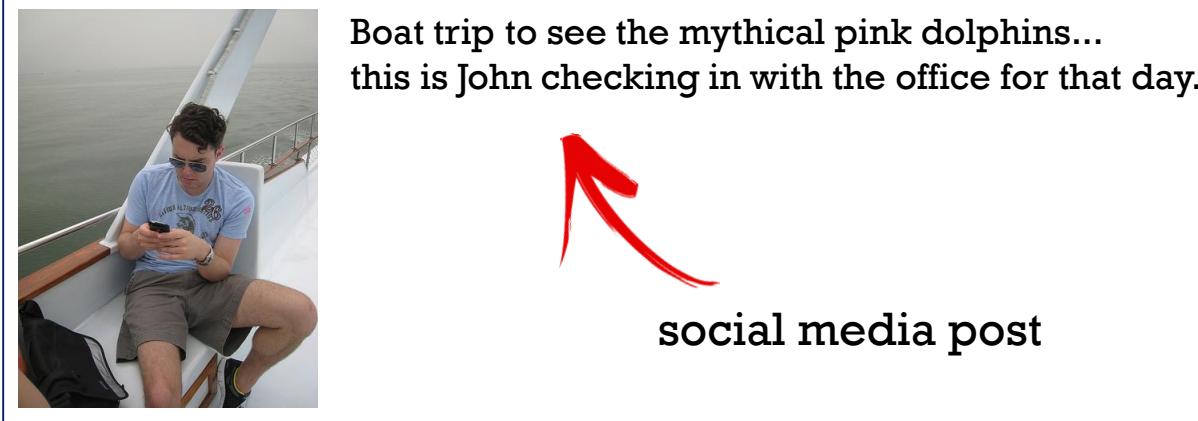
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- ~ 50,000 images with captions

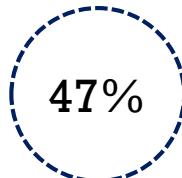
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- Baselines: Text-only and Text + Visual search approaches
- Evaluation metric: Average Precision @ 10

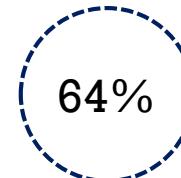
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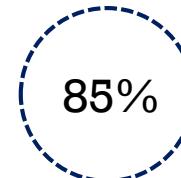
Text-only



Text + Visual



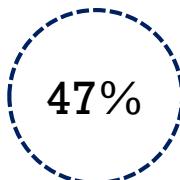
Text + Visual + CSK (Know2Look)



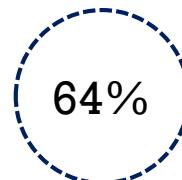
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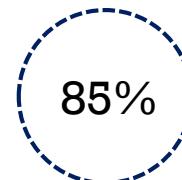
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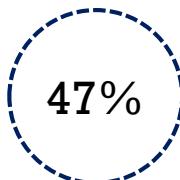
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- Concrete – *ball park, bridge road, table home, bicycle road*
- Abstract – *diesel transport, housing town*
- Mixed – *old clock, backpack travel, boat tour*

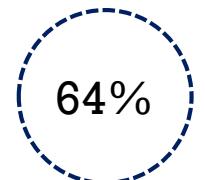
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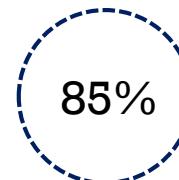
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Co-occurring Flickr tags

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Query: "group excursion"

Text-only



$xx_j$ : "A small excursion boat anchored on the beach at the resort in Mexico."

Text + Visual



$xv_j$ : lunar excursion module, conveyance

Text + Visual + CSK (Know2Look)



$xx_j$ : "A group of people riding camels."  
 $y_k$ : (an excursion, be trip by,  
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- Noisy textual cues + noisy visual object detection + noisy commonsense knowledge → **ensemble effect** → better results for multimodal document retrieval
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Thank you