



# A model for information summarization, and some basic results

**Eric Graves**

ARL

Qiang Ning

UIUC

Prithwish Basu

Raytheon

July 8, 2019

Unclassified



# Presentation Goals



- Derive model for summarizer.
- Present basic results for summarizer.
- Discuss how results are achieved.



# Previous work



## Past Work

- Y. Lin, G. Cao, J. Gao, and J.-Y. Nie, “*An information-theoretic approach to automatic evaluation of summaries*,” in 2006 N.A.A.C.L.-H.L.T.
- H. Lin and J. Bilmes, “*Multi-document summarization via budgeted maximization of submodular functions*,” in 2010 N.A.A.C.L.-H.L.T.
- H. Lin and J. Bilmes, “*Learning mixtures of submodular shells with application to document summarization*,” in 2012 U.A.I.
- A. See, P. J. Liu, and C. D. Manning, “*Get to the point: Summarization with pointer-generator networks*,” in 2017 A.M.A.C.L.
- S. Gehrmann, Y. Deng, and A. Rush, “*Bottom-up abstractive summarization*,” in 2018 E.M.N.L.P.
- ... and many more



# Motivating Example



Phenomena	
High winds	✓
High UV index	✓
Heavy Rain	✓
Snow	
Low visibility	✓
Smog	
Hurricane	✓

## Weather Report



# Motivating Example



Phenomena	
High UV index	✓
Hurricane	✓

## Weather Summary

Phenomena	
High winds	✓
High UV index	✓
Heavy Rain	✓
Snow	
Low visibility	✓
Smog	
Hurricane	✓



# Motivating Example



Phenomena		or, if LA	Phenomena	
High UV index	✓		High UV index	✓
Hurricane	✓		Smog	
			Hurricane	✓

## Weather Summary

Phenomena	
High winds	✓
High UV index	✓
Heavy Rain	✓
Snow	
Low visibility	✓
Smog	
Hurricane	✓



# Motivating Example



Phenomena		or, if LA	Phenomena	
High UV index	✓		High UV index	✓
Hurricane	✓		Smog	

***Is this just compression/rate distortion?***

High UV index	✓
Heavy Rain	✓
Snow	
Low visibility	✓
Smog	
Hurricane	✓



# Motivating Example



Phenomena		or, if LA	Phenomena	
High UV index	✓		High UV index	✓
Hurricane	✓		Smog	

***Is this just compression/rate distortion?***

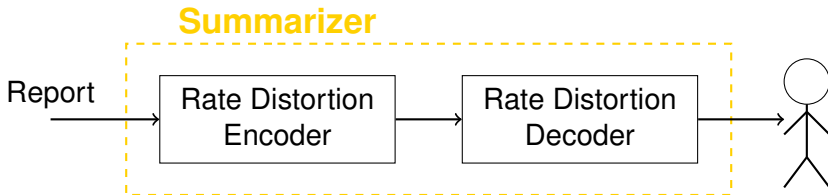
Yes

High UV index	✓
Heavy Rain	✓
Snow	
Low visibility	✓
Smog	
Hurricane	✓





# Model formulation



## Goal

*Minimize  $\mathbb{E}[d(T, S)]$ , for some distortion function\*  $d$ , between the report,  $T$ , and summary,  $S$ , subject to some restrictions on how the summary is generated.*



# Summarizer Restrictions



Phenomena	
High UV index	✓
Smog	✓
Hurricane	✓

⇒ \_1\_\_ \_01

Phenomena	
High winds	✓
High UV index	✓
Heavy Rain	✓
Snow	✓
Low visibility	✓
Smog	✓
Hurricane	✓

⇒ 1110101

Summarizer introduces erasures



# Bad Summary



Phenomena	
High UV index	✓
Smog	⇒ _1___00
Hurricane	

Phenomena		Summarizer <i>only</i> introduces erasures
High winds	✓	
High UV index	✓	
Heavy Rain	✓	⇒ 1110101
Snow		
Low visibility	✓	
Smog		
Hurricane	✓	



# Summarizer Restrictions



Phenomena	
High UV index	✓
Smog	✓
Hurricane	✓

$\Rightarrow \_1\_ \_ \_ 01$

Summarizer *only* introduces erasures

Phenomena	
High winds	✓
High UV index	✓
Heavy Rain	✓
Snow	✓
Low visibility	✓
Smog	✓
Hurricane	✓

$\Rightarrow 1110101$

Summary  $\subset$  Report  
 $S \subset T$

$$\mathcal{T}(s) = \{t : s \subset t\}$$

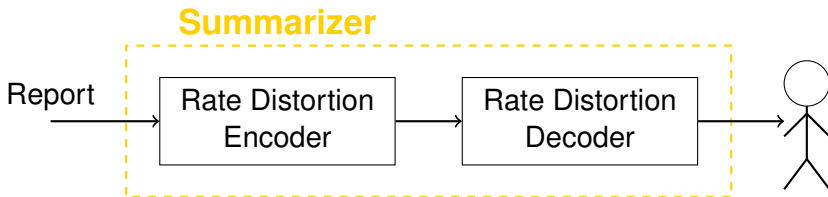


# Model formulation



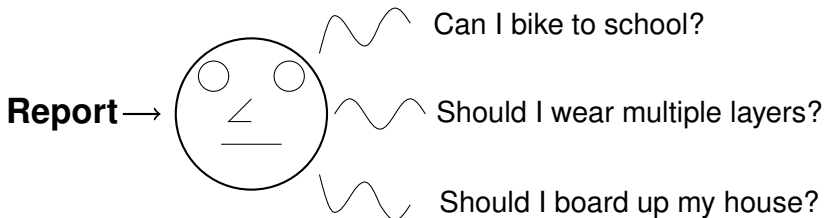
## Goal

Minimize  $\mathbb{E}[d(T, S)]$ , for some distortion function\*  $d$ , between the report,  $T$ , and summary,  $S \subset T$ .





# Distortion function, $d$

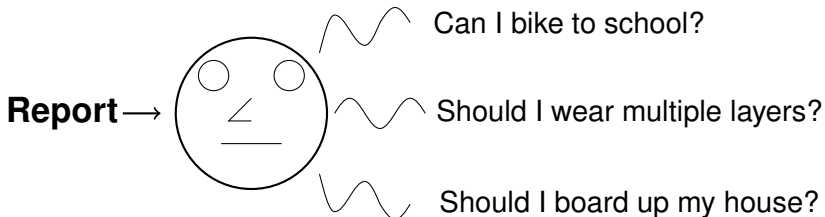




# Distortion function, $d$



**Summaries should answer these questions.**

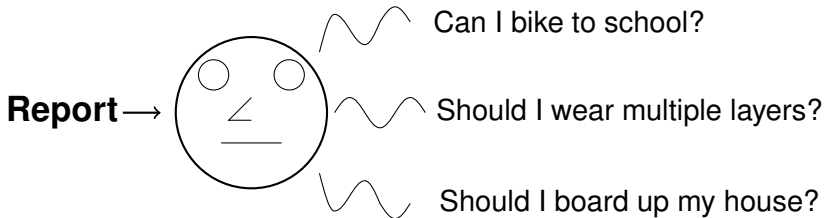




# Assumptions



**Answers are deterministic function of report.**







# Assumptions (Example)



**Answers are deterministic function of report.**

Can I bike to school?

If  $\{\text{Heavy Rain}\} \cup \left\{ \begin{array}{c} \text{High Winds} \\ \text{and} \\ \text{Low visibility} \end{array} \right\} \cup \{\text{Snow}\} \cup \{\text{Hurricane}\}$   
 then **NO.**

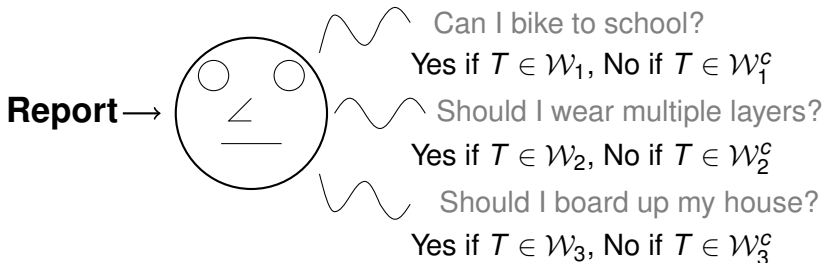
Otherwise **YES.**



# Assumptions



**Answer depends on to which sets the report belongs.**

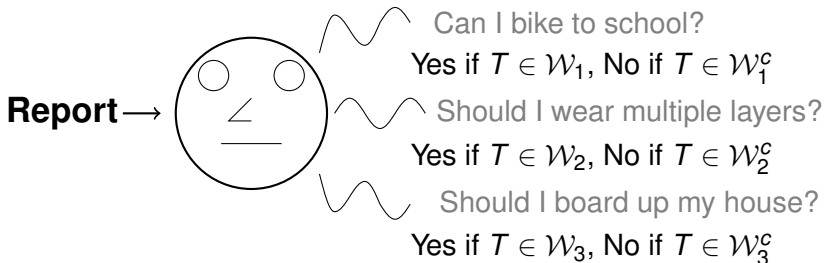




# Model construction



**Cost  $c(t, \mathcal{W})$  represents importance summary  
convey  $t \in \mathcal{W}$ .**

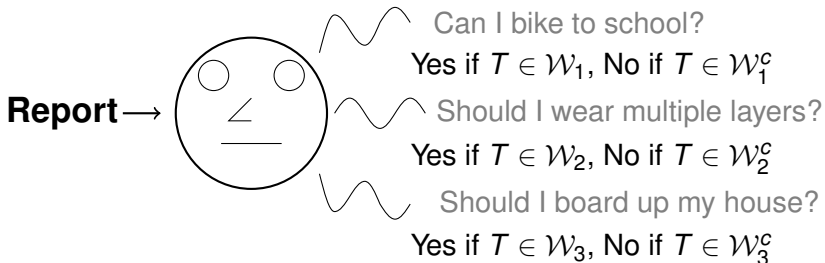




# Model construction



$$d(t, s) = \sum_{\mathcal{W}: t \in \mathcal{W}} c(t, \mathcal{W}) \cdot \text{ambiguity penalty}(\mathcal{W}|s)$$

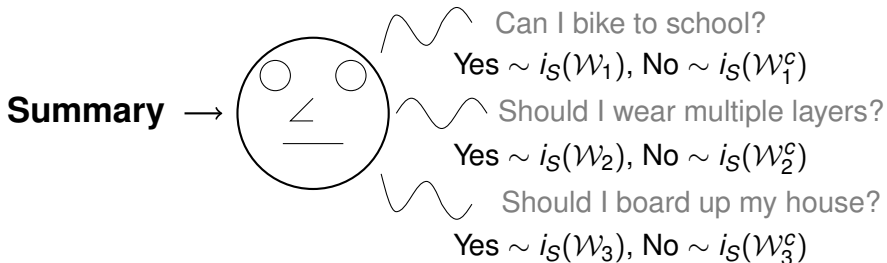




# Ambiguity penalty



$i_s(\cdot)$  models user's estimated probability of report  
· given summary  $s$ .



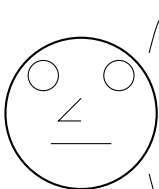


# Ambiguity penalty



$$i_S(\mathcal{W}) = \Pr(T \in \mathcal{W} | \mathbf{s} \subset T) = \frac{p_T(\mathcal{W} \cap \mathcal{T}(\mathbf{s}))}{p_T(\mathcal{T}(\mathbf{s}))}$$

**Summary** →



Can I bike to school?

Yes  $\sim i_S(\mathcal{W}_1)$ , No  $\sim i_S(\mathcal{W}_1^c)$

Should I wear multiple layers?

Yes  $\sim i_S(\mathcal{W}_2)$ , No  $\sim i_S(\mathcal{W}_2^c)$

Should I board up my house?

Yes  $\sim i_S(\mathcal{W}_3)$ , No  $\sim i_S(\mathcal{W}_3^c)$



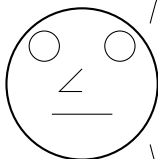
# Ambiguity penalty



Use  $f$ -divergence to measure distance, variational distance gives

$$\text{ambiguity penalty}(\mathcal{W}|s) = 1 - i_s(\mathcal{W})$$

Summary →



Can I bike to school?

Yes  $\sim i_s(\mathcal{W}_1)$ , No  $\sim i_s(\mathcal{W}_1^c)$

Should I wear multiple layers?

Yes  $\sim i_s(\mathcal{W}_2)$ , No  $\sim i_s(\mathcal{W}_2^c)$

Should I board up my house?

Yes  $\sim i_s(\mathcal{W}_3)$ , No  $\sim i_s(\mathcal{W}_3^c)$

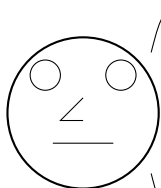


# Model construction



$$\begin{aligned}
 d(t, s) &= \sum_{\mathcal{W}: t \in \mathcal{W}} c(t, \mathcal{W})(1 - i_s(\mathcal{W})) \\
 &= \sum_{\mathcal{W}: t \in \mathcal{W}} c(t, \mathcal{W}) \left( 1 - \frac{p_T(\mathcal{W} \cap \mathcal{T}(s))}{p_T(\mathcal{T}(s))} \right)
 \end{aligned}$$

**Summary** →



Can I bike to school?

Yes  $\sim i_s(\mathcal{W}_1)$ , No  $\sim i_s(\mathcal{W}_1^c)$

Should I wear multiple layers?

Yes  $\sim i_s(\mathcal{W}_2)$ , No  $\sim i_s(\mathcal{W}_2^c)$

Should I board up my house?

Yes  $\sim i_s(\mathcal{W}_3)$ , No  $\sim i_s(\mathcal{W}_3^c)$





# Model



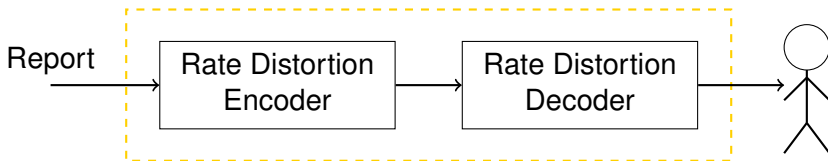
## Goal

Minimize  $\mathbb{E}[d(T, S)]$ , where

$$d(t, s) = \sum_{W:t \in W} c(t, W) \left( 1 - \frac{p_T(W \cap \mathcal{T}(s))}{p_T(\mathcal{T}(s))} \right),$$

between the report,  $T$ , and summary,  $S \subset T$ .

## Summarizer





# Model



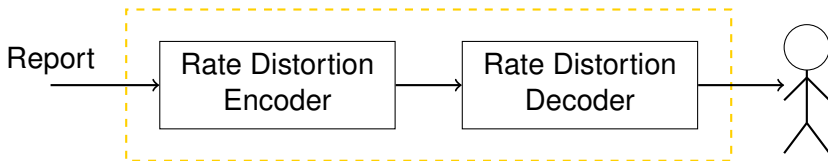
## Goal

Minimize  $\mathbb{E}[d(T, S)]$ , where

$$d(t, s) = \sum_{W: t \in W} c(t, W) \left( 1 - \frac{p_T(W \cap \mathcal{T}(s))}{p_T(\mathcal{T}(s))} \right),$$

between the report,  $T$ , and summary,  $S \subset T$ .

## Summarizer





# Model



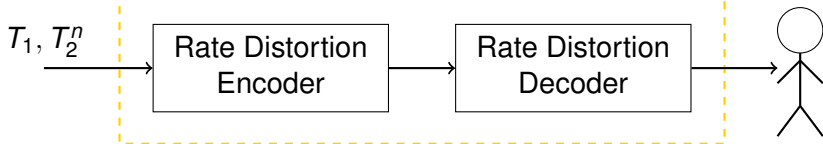
## Goal (Universal Summarization)

Minimize

$$\int_{\mathcal{P}(\mathcal{T})} \sum_{t^n} \prod_{i=1}^n p_{\mathcal{T}}(t_i) \sum_{\mathcal{W}: t_1 \in \mathcal{W}} c(t_1, \mathcal{W}) \left( 1 - \frac{p_{\mathcal{T}}(\mathcal{W} \cap \mathcal{T}(s))}{p_{\mathcal{T}}(\mathcal{T}(s))} \right) dr$$

where  $\frac{dr}{dp_{\mathcal{T}}}$  is constant and  $\int_{\mathcal{P}(\mathcal{T})} dr = 1$ .

## Summarizer





# Results



## Theorem (Optimal universal summarizer)

The optimal summarizer chooses the summary  $s(t_1 | t_2^n)$  that minimizes

$$\sum_{\mathcal{W}: t_1 \in \mathcal{W}} c(t_1, \mathcal{W}) [1 - q(\mathcal{W} \cap \mathcal{T}(s))] \eta_{t^n, s}$$

where

$$\eta_{t^n, s} = \sum_{k=0}^{\infty} \frac{(n + |\mathcal{T}| + k - \pi(\mathcal{T}(s) | t^n) - 2)! (n + |\mathcal{T}|)!}{(n + |\mathcal{T}| + k)! (n + |\mathcal{T}| - \pi(\mathcal{T}(s) | t^n) - 2)!},$$

$$q(a) = \frac{\pi(a | t^n) + 1}{n + |\mathcal{T}|}. \quad (1)$$



# Analysis



## Proof.

Calculate

$$\int_{\mathcal{P}(\mathcal{T})} \prod_{i=1}^n p_{\mathcal{T}}(t_i) \left( 1 - \frac{p_{\mathcal{T}}(\mathcal{W} \cap \mathcal{T}(s))}{p_{\mathcal{T}}(\mathcal{T}(s))} \right) dr,$$

by writing  $p_{\mathcal{T}}$  as a  $|\mathcal{T}|$ -dimensional vector, using Taylor's theorem, and then applying  $\int_0^y (y-x)^a x^b dx = \frac{a!b!}{(a+b+1)!} y^{a+b+1}$  recursively.



## Goal (Universal Summarization)

Minimize

$$\int_{\mathcal{P}(\mathcal{T})} \sum_{t^n} \prod_{i=1}^n p_{\mathcal{T}}(t_i) \sum_{\mathcal{W}: t_i \in \mathcal{W}} c(t_i, \mathcal{W}) \left( 1 - \frac{p_{\mathcal{T}}(\mathcal{W} \cap \mathcal{T}(s))}{p_{\mathcal{T}}(\mathcal{T}(s))} \right) dr$$

where  $\frac{dr}{dp_{\mathcal{T}}}$  is constant and  $\int_{\mathcal{P}(\mathcal{T})} dr = 1$ .



# Results



## Lemma

For positive integers  $b$ ,  $a$  such that  $1 \leq b < b + 2 \leq a$ ,

$$\frac{a+1}{a-b} < \sum_{y=0}^{\infty} \frac{(b+y)!a!}{(a+y)!b!} \leq \frac{a+1}{a-b} (1 + \varepsilon(a-b))$$

where

$$\varepsilon(x) = 3 \frac{1 + \ln(x)}{x} + 4e^{\frac{1}{12}} \cdot 2^{-x/2}.$$

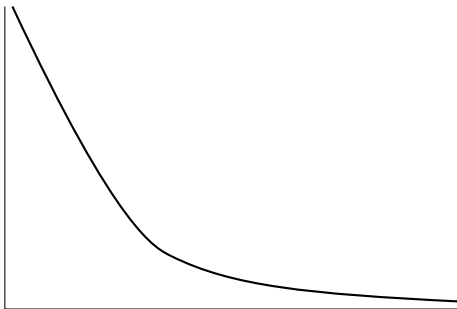


# Analysis



Proof.

$$\text{Plotting } y = \frac{(b+x)!a!}{(a+x)!b!}$$



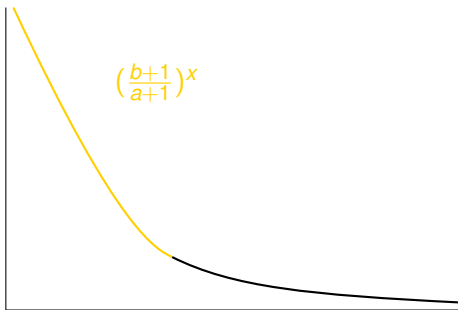


# Analysis



Proof.

$$\text{Plotting } y = \frac{(b+x)!a!}{(a+x)!b!}$$





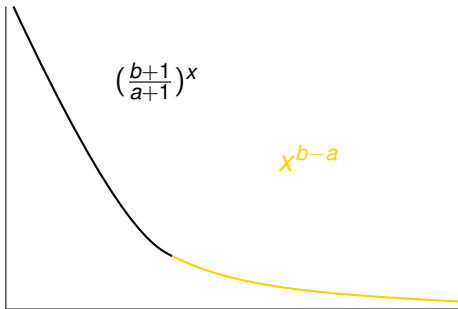


# Analysis



Proof.

$$\text{Plotting } y = \frac{(b+x)!a!}{(a+x)!b!}$$





# Best results



## Theorem

The (nearly) optimal summarizer chooses the summary  $s(t_1|t_2^n)$  that minimizes

$$\sum_{\mathcal{W}: t_1 \in \mathcal{W}} c(t_1, \mathcal{W}) \left[ 1 - \frac{q(\mathcal{W} \cap \mathcal{T}(s))}{\hat{q}(\mathcal{T}(s))} \right]$$

where

$$q(a) = \frac{\pi(a|t^n) + 1}{n + |\mathcal{T}|}, \quad \hat{q}(a) = \begin{cases} \frac{\pi(a|t^n) + 2}{n + |\mathcal{T}| + 1} & \text{if } a = t_1 \\ \frac{\pi(a|t^n) + 1}{n + |\mathcal{T}| + 1} & \text{else} \end{cases} .$$



# Conclusions



Any questions?