

# http://warp.whoi.edu/co-robotic-exploration/

This project is focused on developing new techniques to enable interactive exploration in unknown, low bandwidth environments, with a multi-robot team. The proposed approach enables new types of



# lem

m to develop teams of under oration robots capable of cateeabed terrains. This is chal

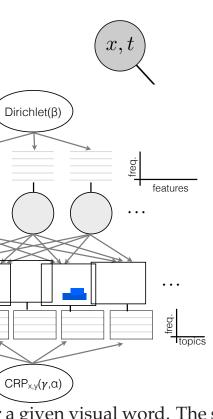
supervised models like topic els lack definitive "cor ndence" between categories

nmunication-constraint lerwater prevent high dwidth solutions (e.g. transsion of raw images)

improved consistency becal topic models with limited

# iotemporal Topic Models

s on unsupervised categorization using Bayesian nonparametric infere use a "bag of words" model, where images are represented by a set of words", i.e. image features.



Word Dis

odel of the ROST framework

ral topic modeling (ROST) [1] to learn a set probability of word (i.e. image feature)  $w_i$ time *t* as follows:

$$(w_i \mid z_i = k)p(z_i = k \mid x, t)$$
(4)

the posterior probability of a topic assigna given visual word. The sampling distribution is computed as follows:

$$= k \mid \mathbf{z}_{-i}, \mathbf{w}) \propto \left[ \frac{n_{k,-i}^{(v)} + \beta}{n_{k,-i}^{(\cdot)} + V\beta} \right] \left[ \frac{n_k^{(G_i)} + \alpha}{n_{-i}^{(G_i)} + K\alpha} \right],$$
(5)

is the count of assignments of topic k to every other observation of  $n_{k-i}^{(\cdot)}$  is the count of all current assignments of topic k,  $n_k^{(G_i)}$  is the count signments of topic k in spatiotemporal neighborhood  $G_i$ , and  $n_{-i}^{(G_i)}$  is the onding total count of topic assignments to all other words in  $G_i$ . mum-likelihood estimates for topic-word distributions and topic mixing ons respectively can then be computed as follows:

der

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Br

$$\hat{\phi}_{kw} = \frac{n_k^{(w)} + \beta}{n_k^{(\cdot)} + V\beta}, \ \hat{\theta}_{G_ik} = \frac{n_k^{(G_i)} + \alpha}{n^{(G_i)} + K\alpha}.$$
(6)

ecologists.

sparse, and previously.

in extreme

environments-like the deep<sup>acoustic</sup>ea.



Figure 1: Two AUV's must merge local topic models via a central node (ship).

these descriptors, given by:

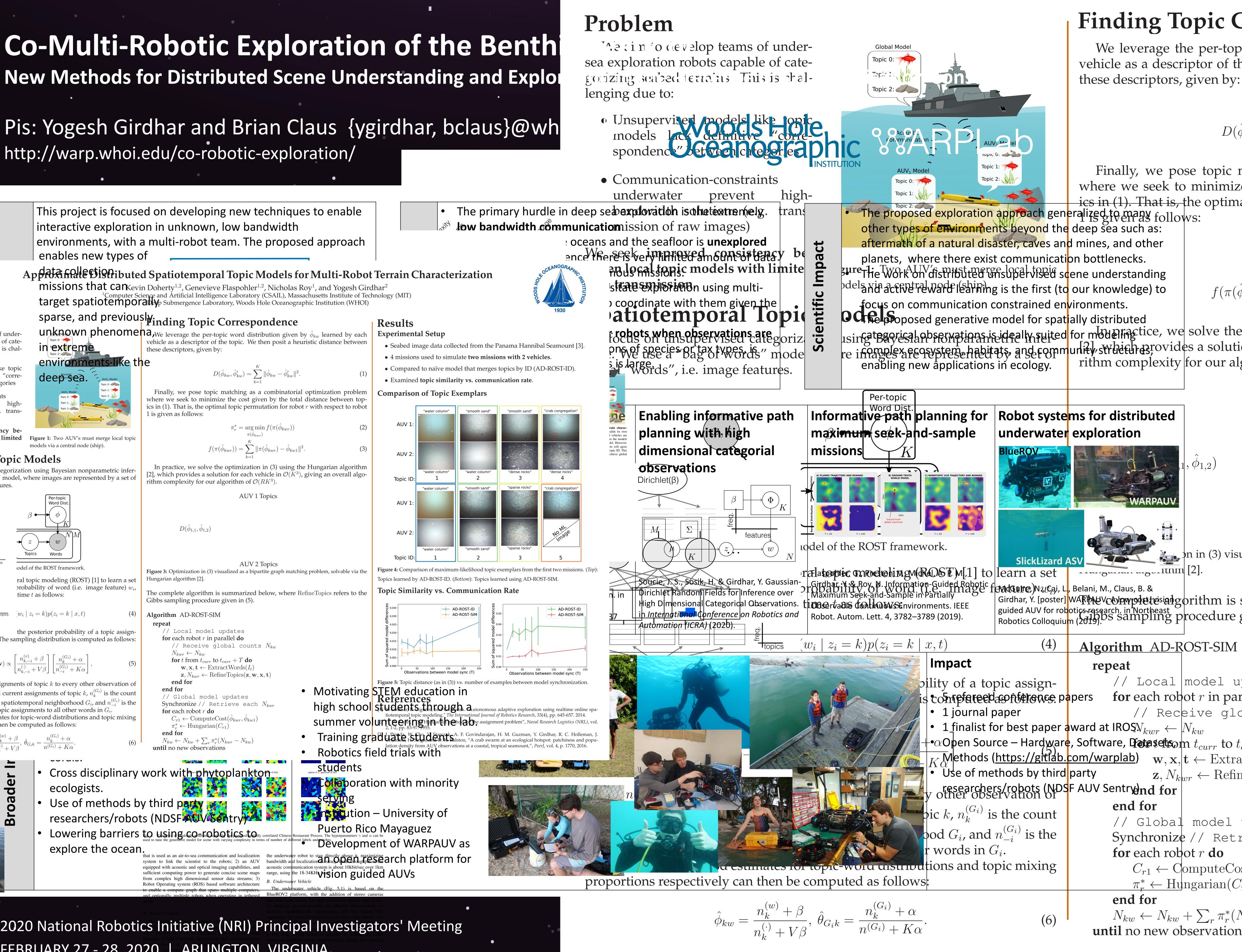
$$D(\hat{\phi}_{kw}, \hat{\phi}'_{kw}) = \sum_{k=1}^{K} \|\hat{\phi}_{kw} - \hat{\phi}'_{kw}\|^2.$$

$$\pi_r^* = \underset{\pi(\hat{\phi}_{kwr})}{\arg\min} f(\pi(\hat{\phi}_{kwr}))$$
$$f(\pi(\hat{\phi}_{kwr})) = \sum_{k=1}^K \|\pi(\hat{\phi}_{kwr}) - \hat{\phi}_{kw1}\|^2.$$

rithm complexity for our algorithm of  $\mathcal{O}(RK^3)$ .

**AUV 2 Topics** 

Gibbs sampling procedure given in (5).



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