Adaptive Methane Detection

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Project Overview

Simulate Methane Distribution

Update Hypothesis Probability

Expected Information Gain to start path planning

Project Object

- JPL has developed a hand-held methane sniffer that can be deployed on a flying robot.
- To develop an automated system that guides the robot or person carrying the sniffer from first detection to the leak source.
- Input: methane concentration and wind vector at detected location
- Output: direction to the leak source

Foundation Approach

- The approach is a Bayesian Model
- First, set several hypotheses (leak location, leak concentration)
- Using the collocated data to update the probability of those hypotheses

Project Overview

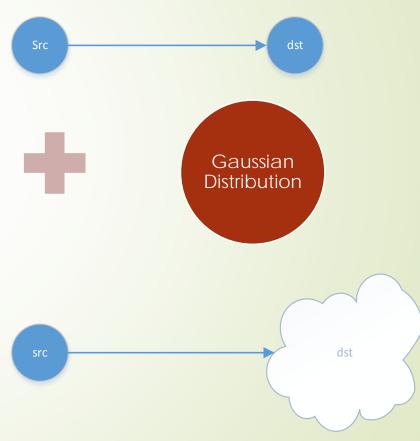
Simulate Methane Distribution

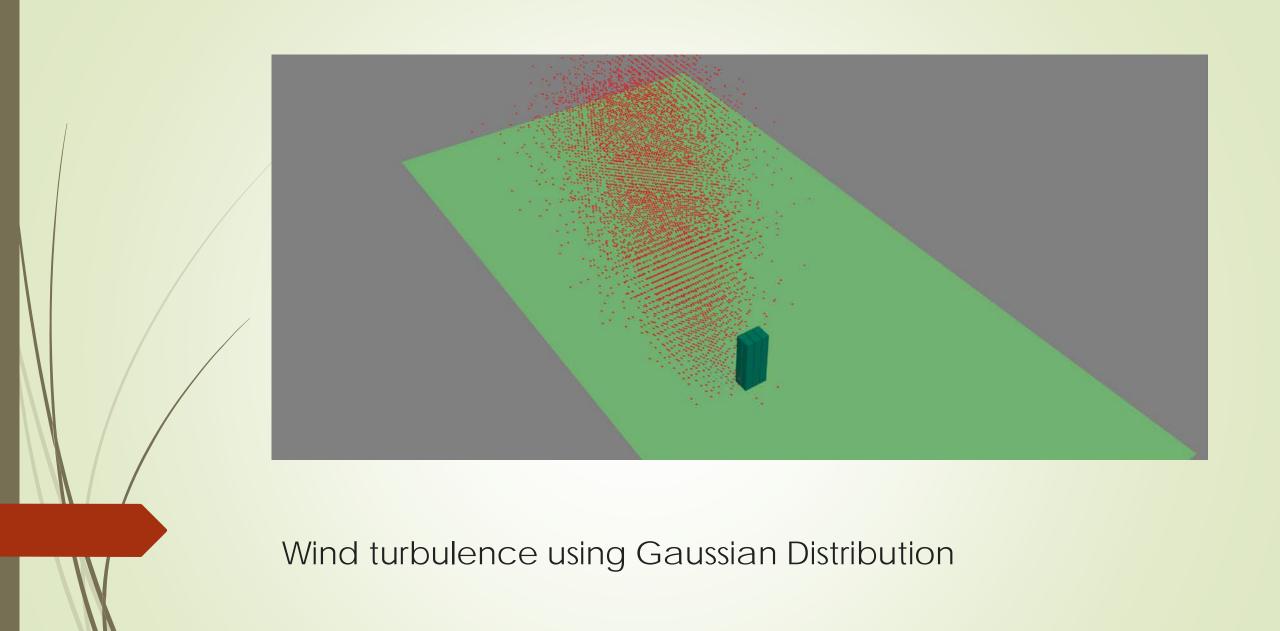
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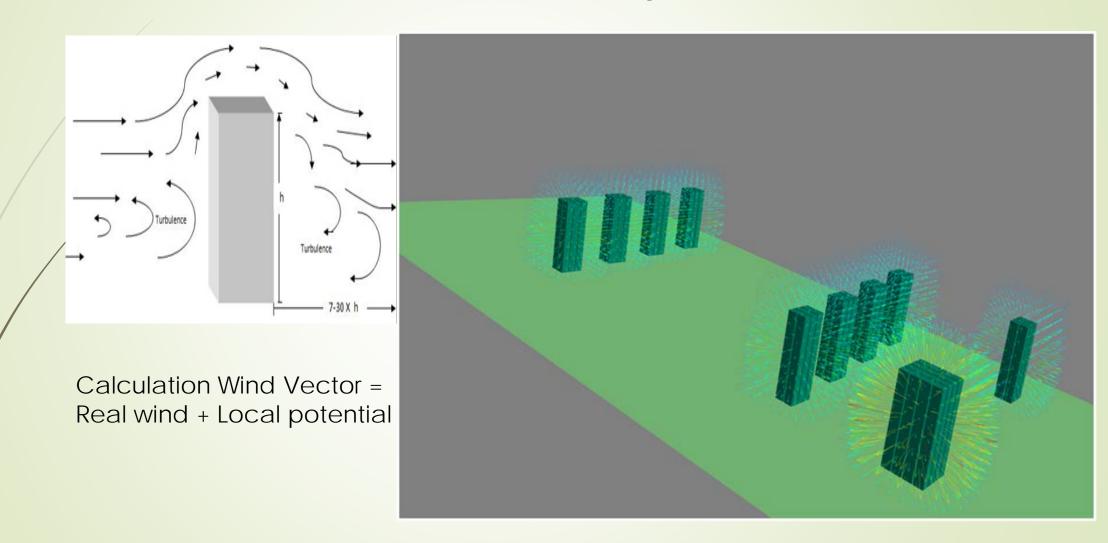
Simulate wind turbulence





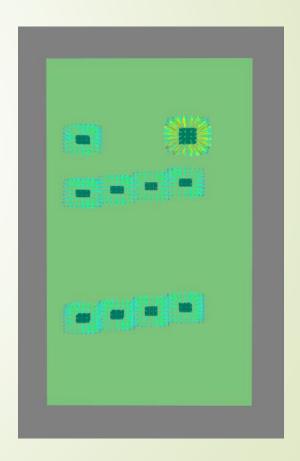


Local potential field by obstacle



Build the map model





Build a virtual world for hypothesis

- Decompose the map into cells tagged as building, air and ground.
- Make a methane leak location cell and its leak concentration as a hypothesis.
- Distribute methane from cells to cells using Gaussian distribution by calculated wind.

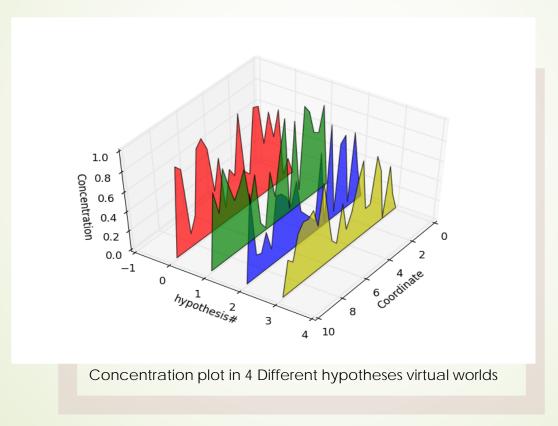
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Calculated methane concentration under each hypothesis



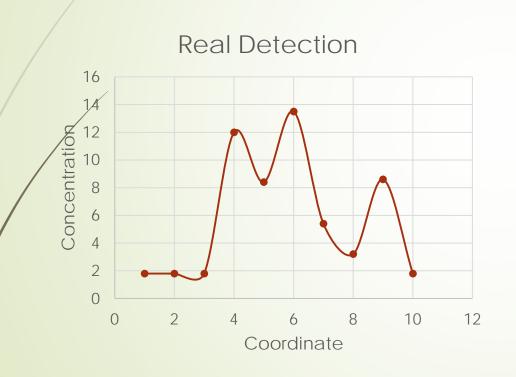
We normalized the probability for all the hypotheses in the initialized stage:

 $probability^i = 1/N$

For the virtual world according to each hypothesis, calculate the methane concentration in every cell in the map.

We can draw the concentration/coordinate plot as left showing

Model the likelihood using gamma distribution









Model the likelihood using gamma distribution

 $Mean^i$ = Concentration under i^{th} hypothesis

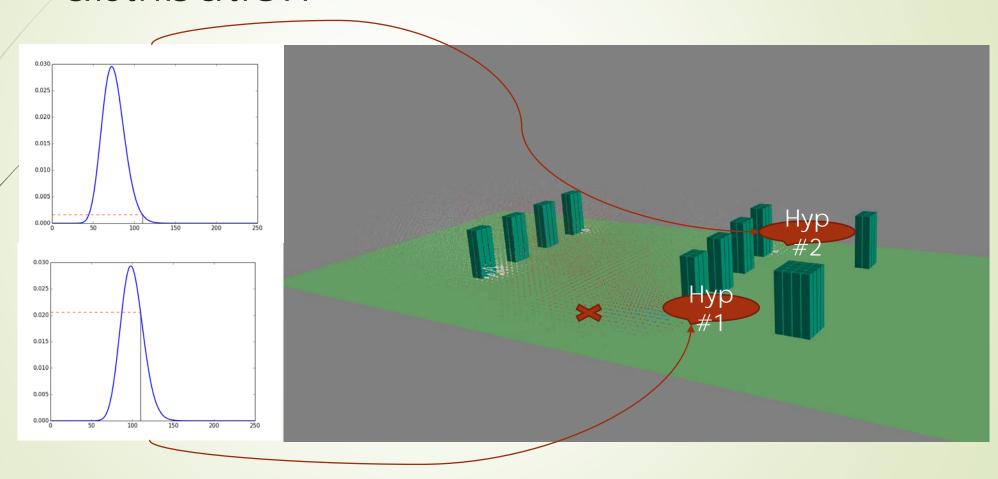
x = Concentration detected in real word

 $likelihood^i = Gamma.pdf(Mean^i, x)$

 $Probability_{new}^i = Probability_{old}^i * likehood^i$

Repeated the steps for each detection

Model the likelihood using gamma distribution



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Entropy for the set of hypotheses

- Make probability of hypothesis as variable marked as xⁱ
- Mark the all hypotheses set as X
- Entropy of the set of hypotheses is:
 - $H(X) = -\sum_{i} P(x^{i}) log_{b} P(x^{i})$

Update probability in the future under specific assumption

- Prepared a location l as detected candidate in the future
- Assume the k^{th} hypothesis is true for next detection
- Calculate new probability for each hypothesis under that assumption:
 - \blacksquare likelihoodⁱ_{k,l} = Gamma.pdf(Meanⁱ_l, Mean^k_l)
 - $ightharpoonup Probability_{k,l}^i = Probability^i * likelihood_{k,l}^i$

Expected Information Gain under specific assumption

The new Entropy for the probability updated hypotheses:

$$H(X_{k,l}) = -\sum_{i} P(x_{k,l}^{i}) * \log_{b} P(x_{k,l}^{i})$$

Expected Information Gain:

$$\blacksquare IG_{k,l} = H(X) - H(X_{k,l})$$

Sum the Expected Information Gain for each assumption

- Assume all the hypotheses are true one hypothesis a time
- Sum all the Expected Information Gain under one true hypothesis assumption weighted by its original probability:

$$\blacksquare IG_l = \sum_k IG_{k,l} * P(x^k)$$

Choose candidates by Expected Information Gain

- By compared IG_l for different locations, we can decide which candidate location to move next step in order to get more information.
- It's a start point for path planning

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A virtualization Demo

- Thank David for all his the intelligent ideas and algorithm framework
- Thank Lance for his real world knowledge and practical experience to build our model
- Here is a Demo running on real data collected by Lance.