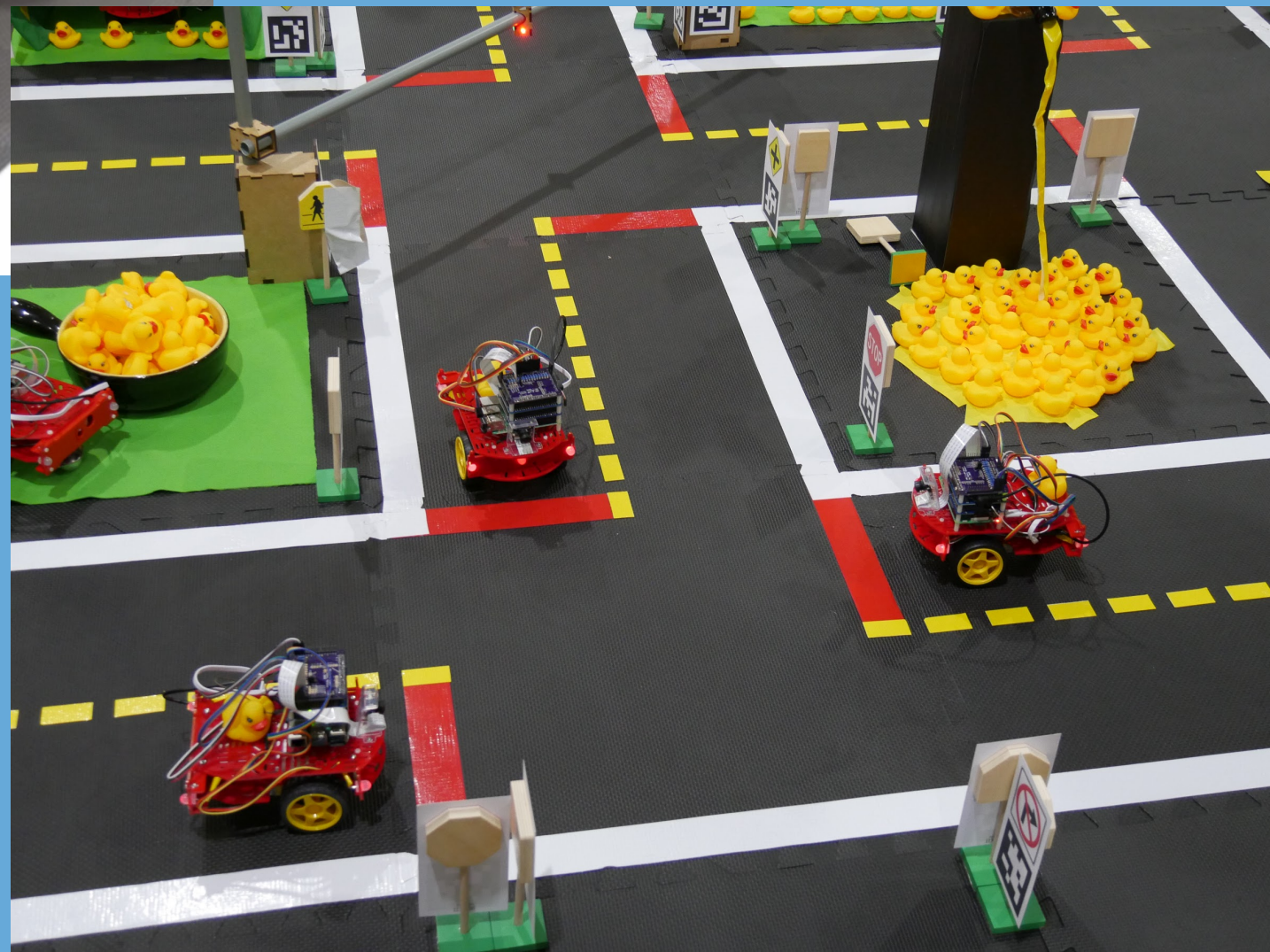


**CS 3630!**



***Lecture 10:  
Continuous Densities***

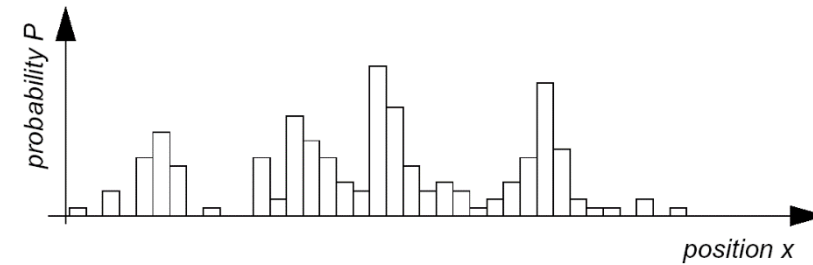


# Topics

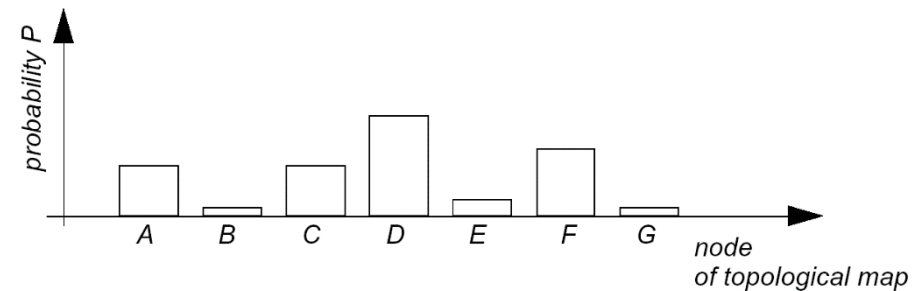
- **1. Continuous Densities**
- **2. Gaussian Densities**
- **3. Bayes Nets & Mixture Models**
- **4. Cont. Measurement Models**
- **5. Cont. Motion Models**
- **6. Simulating Cont. Bayes Nets**

Motivation: how do we represent our belief of where the robot is located?

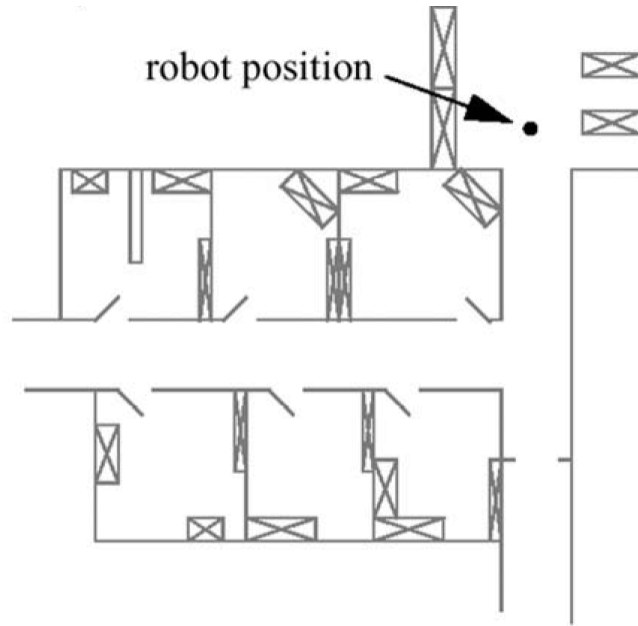
Discretized map with multiple hypotheses probability distribution



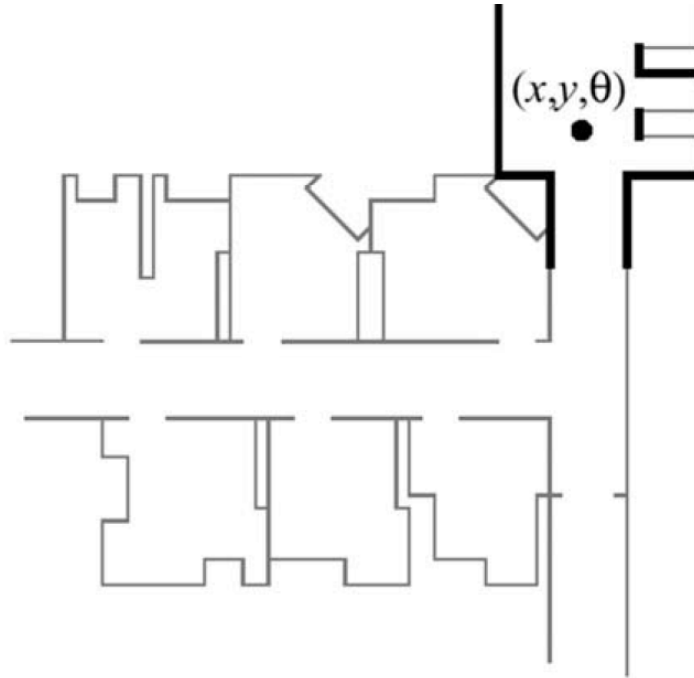
Discretized topological map with multiple hypotheses probability distribution



# Representations

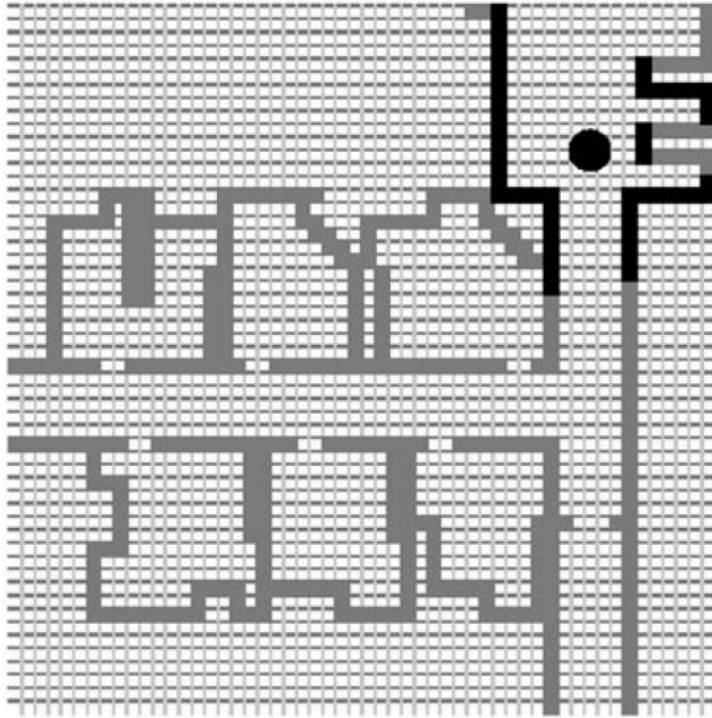


The real map with walls, doors and furniture.

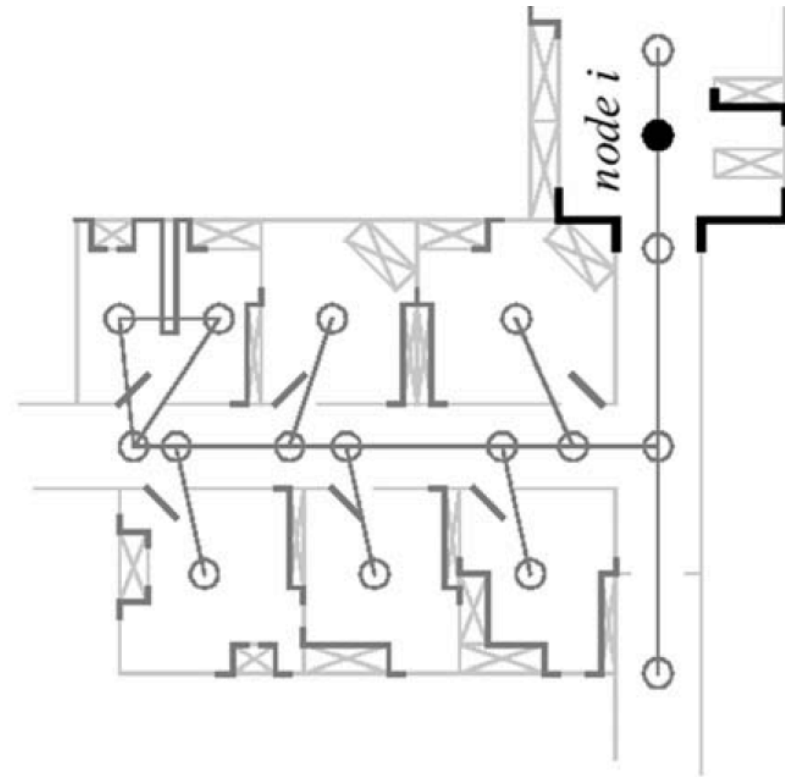


Line-based map (~100 lines)

# Representations



Grid-based map (3000 cells,  
each 50cm x 50cm)



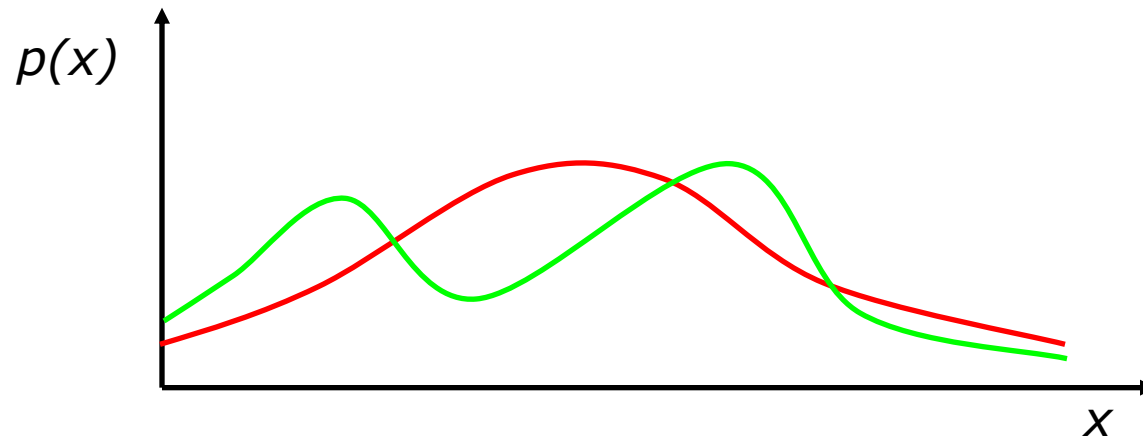
Topological map (50 features,  
18 nodes)

# 1. Continuous Probability Densities

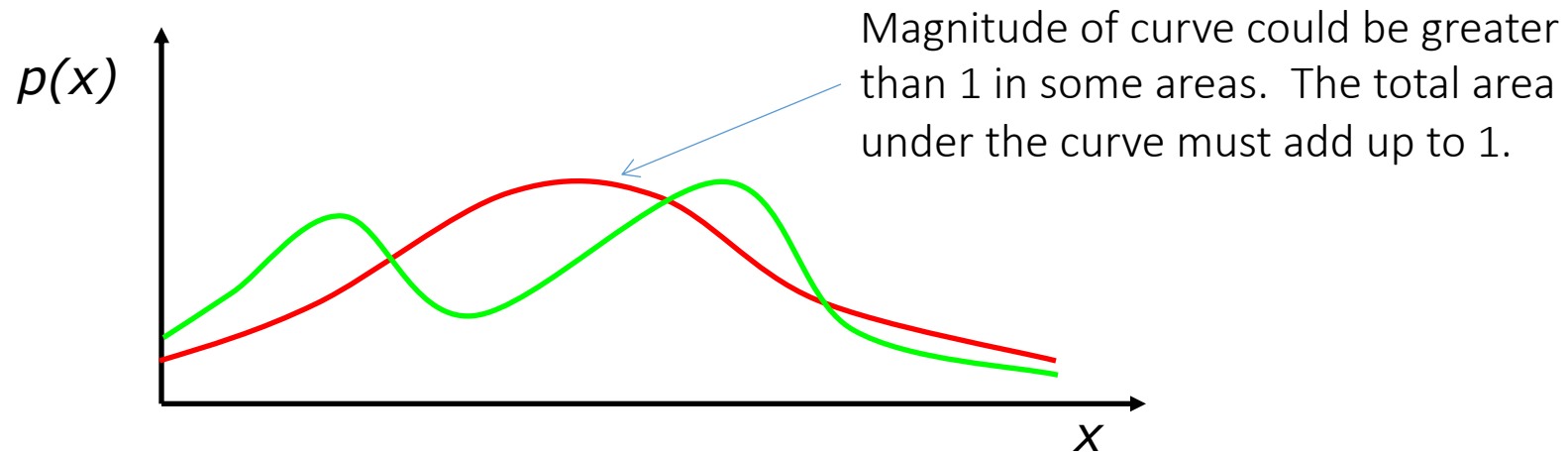
- $X$  takes on values in the continuum.
- $p(X = x)$ , or  $p(x)$ , is a **probability density function**.

$$P(x \in (a, b)) = \int_a^b p(x) dx$$

- E.g.



# Probability Density Function



Since continuous probability functions are defined for an infinite number of points over a continuous interval, the probability at a single point is always 0.

## 2. Gaussian Densities

A Gaussian probability density is given by

$$\mathcal{N}(\theta; \mu, \Sigma) = \frac{1}{\sqrt{|2\pi\Sigma|}} \exp \left\{ -\frac{1}{2} \|\theta - \mu\|_{\Sigma}^2 \right\},$$

where  $\mu \in \mathbb{R}^n$  is the mean,  $\Sigma$  is an  $n \times n$  covariance matrix, and

$$\|\theta - \mu\|_{\Sigma}^2 \triangleq (\theta - \mu)^{\top} \Sigma^{-1} (\theta - \mu)$$

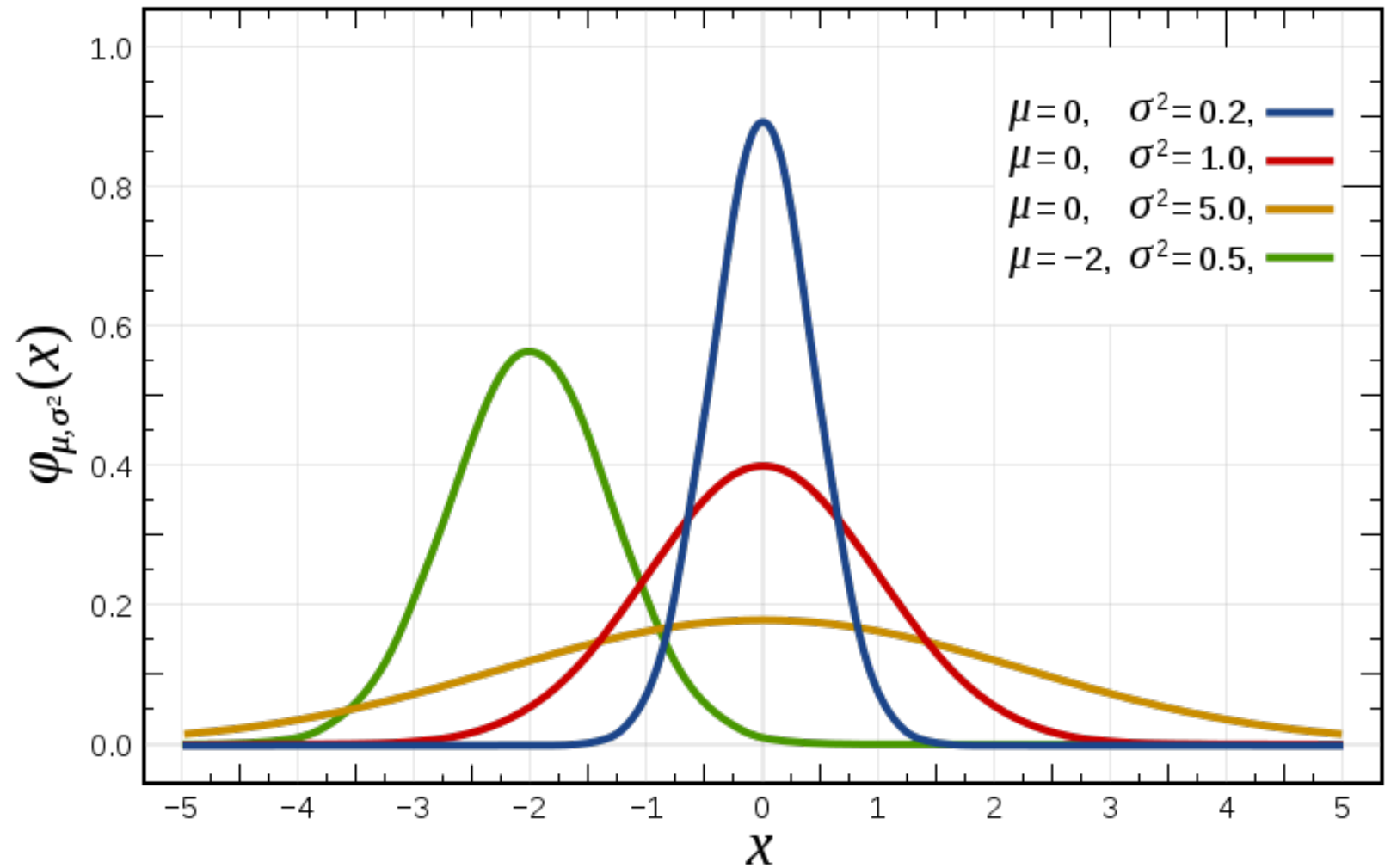
denotes the squared Mahalanobis distance.

- Easy: negative log is quadratic
- Also known as the “bell curve”
- One of a few densities for which sampling is *easy*

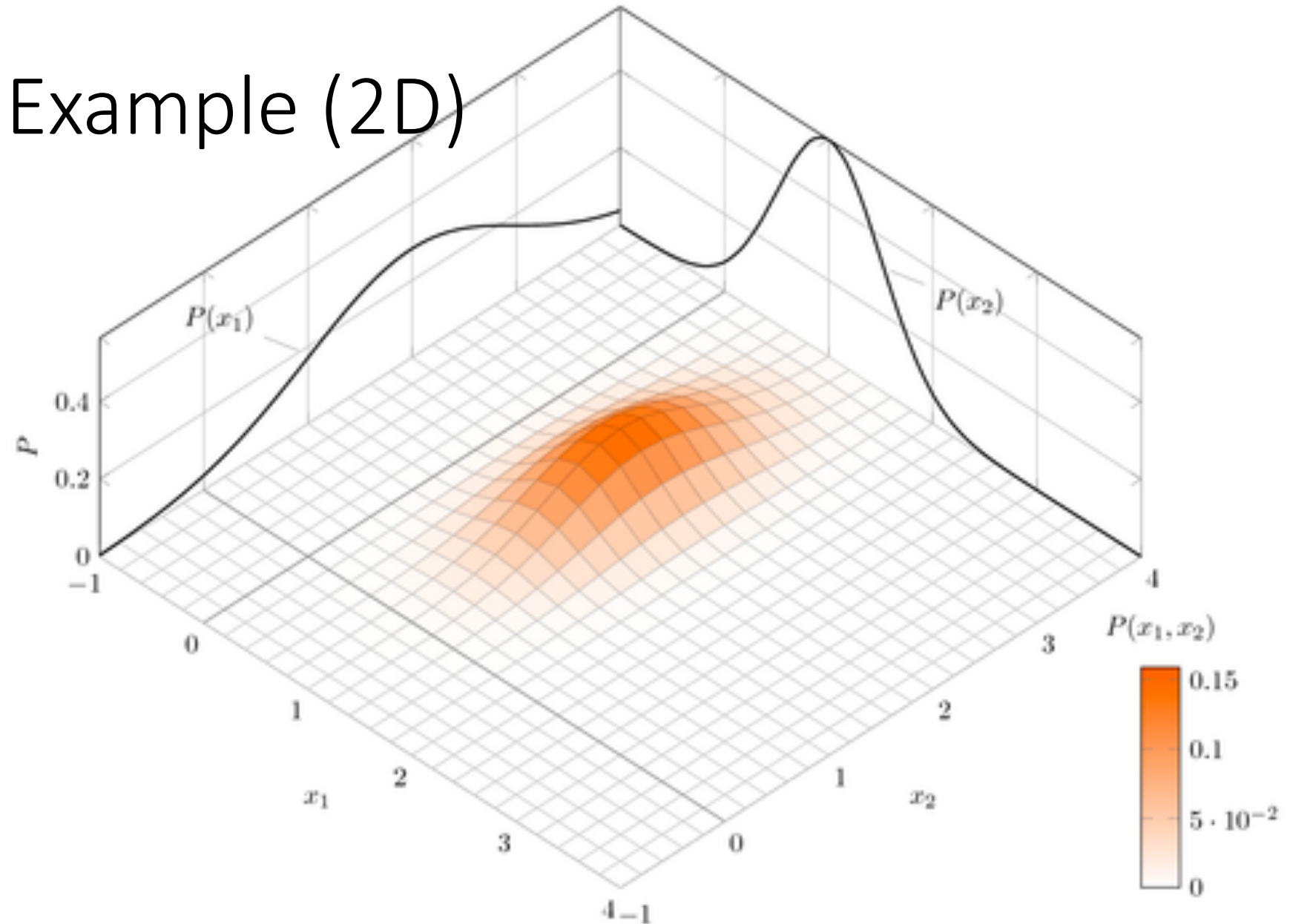


# 1D examples

- From Wikipedia!



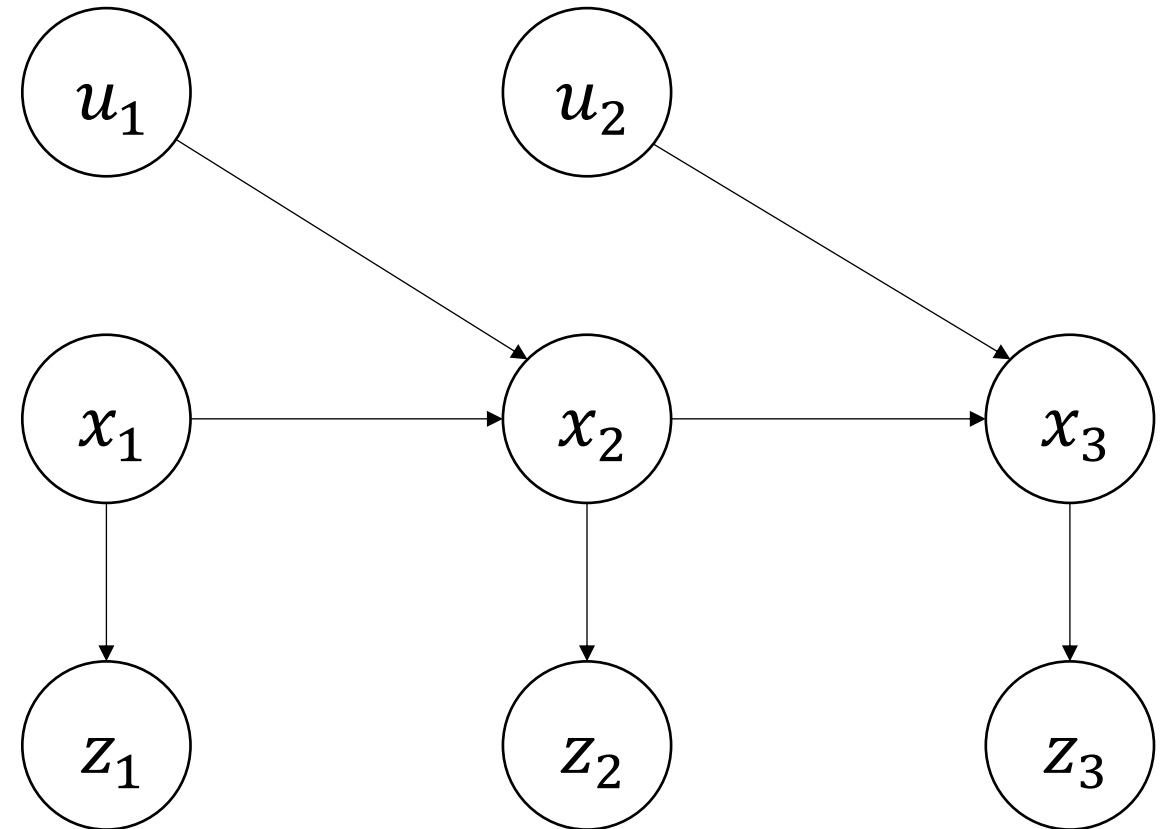
# Multivariate Example (2D)



- <http://pgfplots.net/tikz/examples/bivariate-normal-distribution/>

## 2. Continuous Bayes Nets

- As before, but now states  $S$ , observations  $O$ , and action  $A$  can all be continuous.
- Terminology:  $x$ ,  $z$ ,  $u$
- Hence: measurement models and state transition models are continuous.

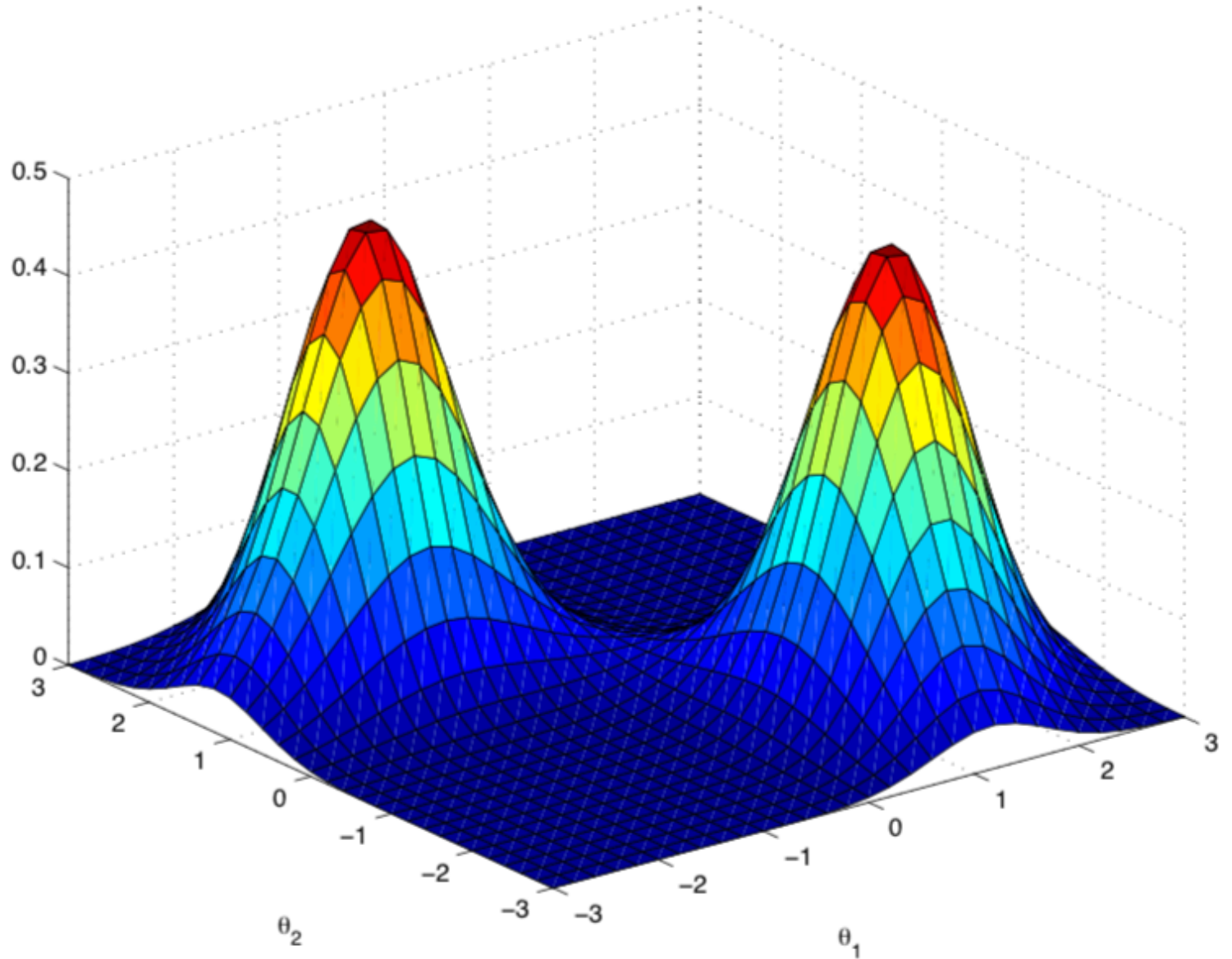


# Important aside: Mixture Models



- We can mix discrete and continuous
- Most important example: mixture of continuous densities
- Example: Gaussian mixture model
- Sampling: sample **component**, then sample from Gaussian:

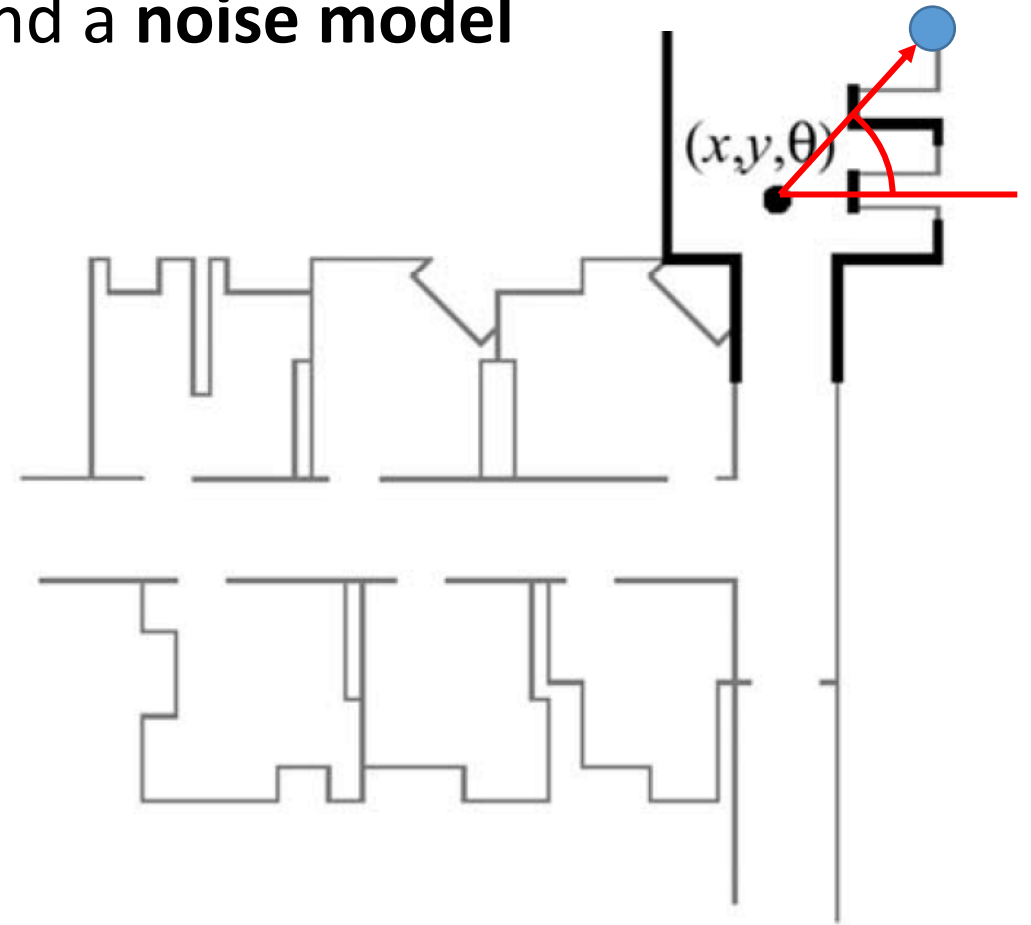
$$p(x, C) = p(x|C)P(C)$$



## 4. Continuous Measurement Models

- We need a **measurement function** and a **noise model**
- Example: bearing to a landmark  $l$ :

$$h(x, l) = \text{atan2}(l_y - x_y, l_x - x_x)$$



# Adding a noise model

- Generative model of measurement  $z = h(x, l) + \eta$ ,
- Assuming Gaussian noise:

$$p(z|x, l) = \mathcal{N}(z; h(x, l), R) = \frac{1}{\sqrt{|2\pi R|}} \exp \left\{ -\frac{1}{2} \|h(x, l) - z\|_R^2 \right\}$$

# Adding a noise model

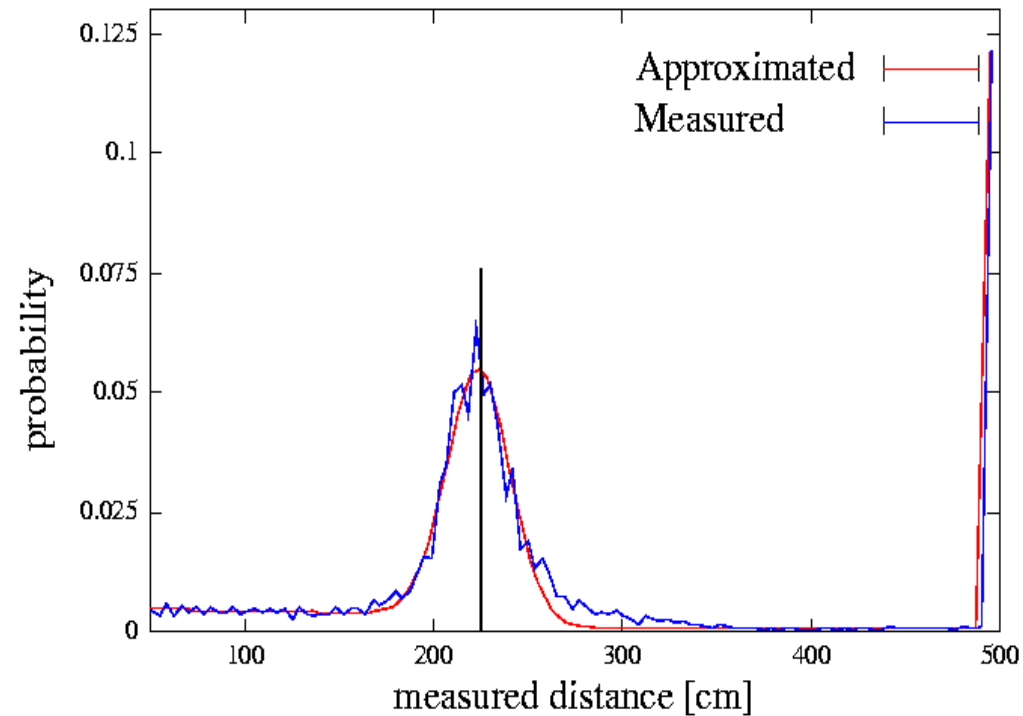
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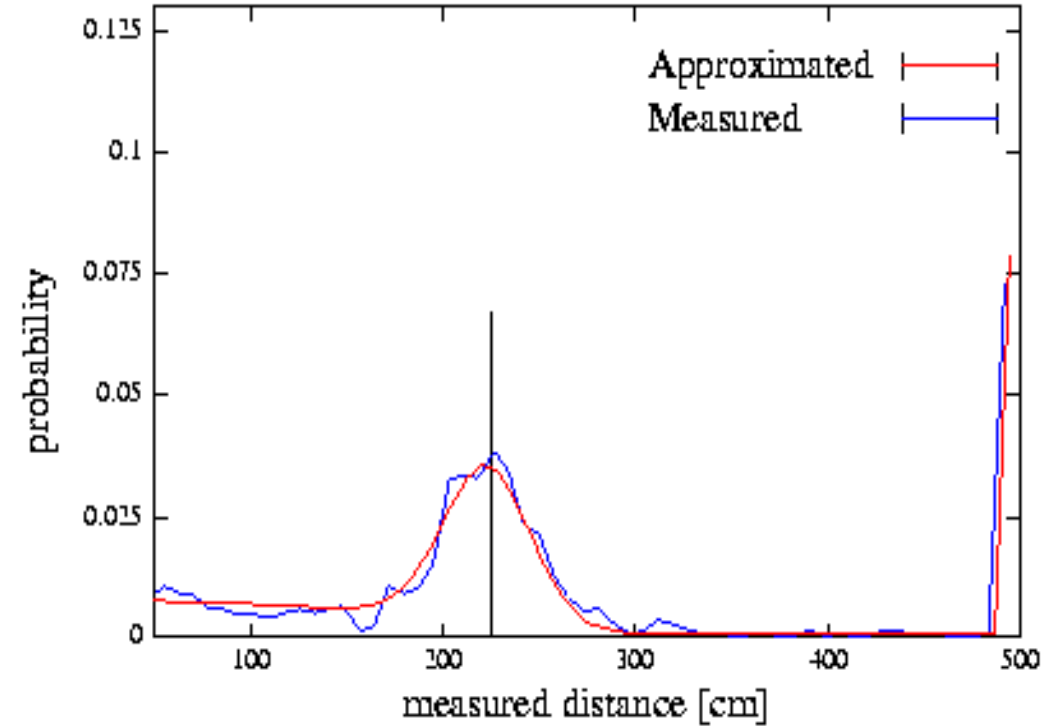
- Putting it together:

$$p(z|x, l) = \frac{1}{\sqrt{|2\pi R|}} \exp \left\{ -\frac{1}{2} \|\text{atan2}(l_y - x_y, l_x - x_x) - z\|_R^2 \right\}$$

# Other sensor models



**Laser sensor**



**Sonar sensor**



## 5. Continuous Motion Models

- Similar for state transition, but we now have a motion model
- Motion model  $g(x, u)$  takes state  $x$  and control  $u$
- Multivariate noise model with covariance  $Q$ :

$$p(x_{t+1}|x_t, u_t) = \frac{1}{\sqrt{|2\pi Q|}} \exp \left\{ -\frac{1}{2} \|g(x_t, u_t) - x_{t+1}\|_Q^2 \right\}$$

# 6. Simulating from a Continuous Bayes Net

## 1. Slice 1:

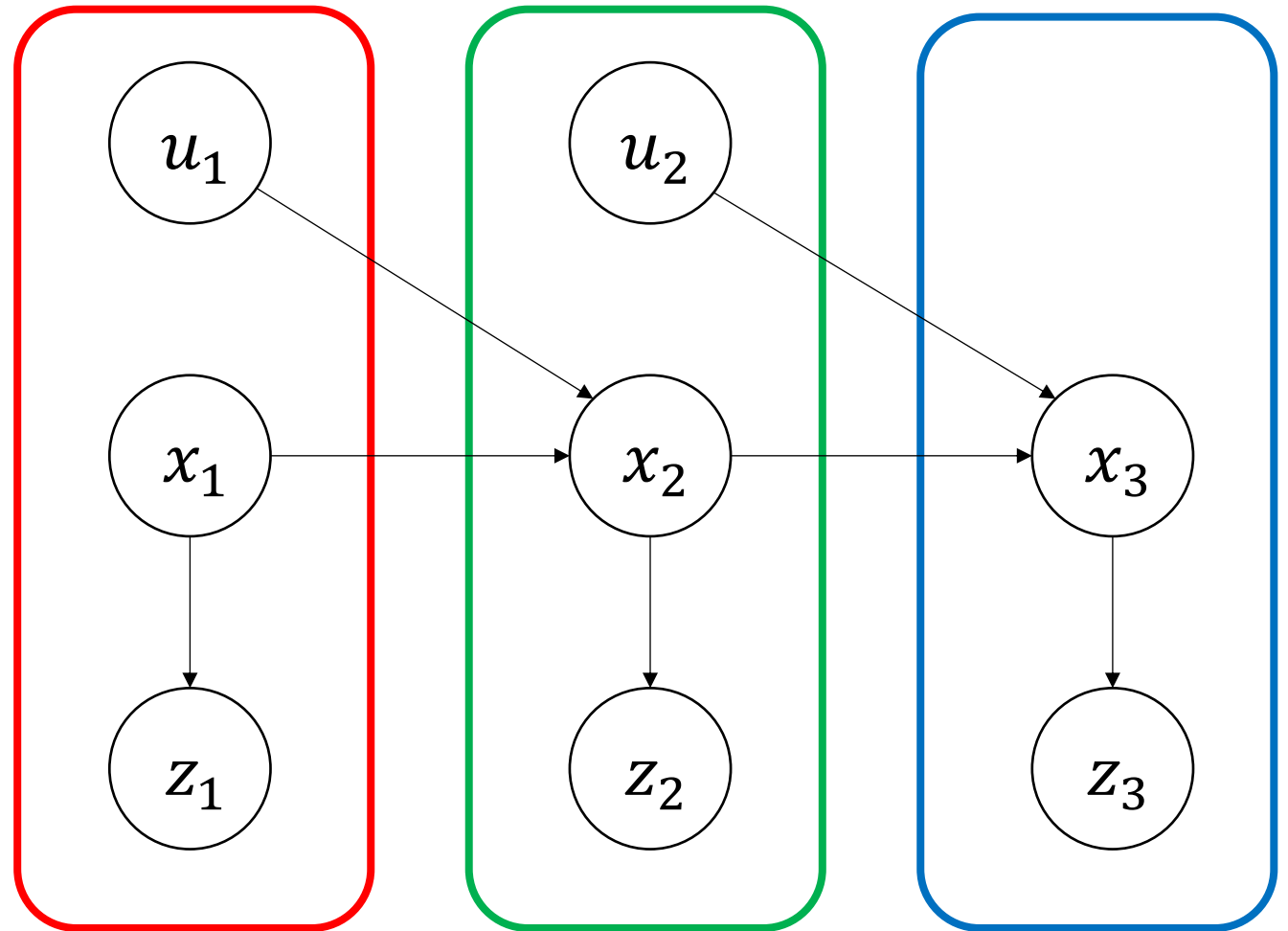
- a) Sample from  $p(x_1)$
- b) Sense  $p(z_1|x_1)$
- c) Sample from  $p(u_1)$

## 2. Slice 2:

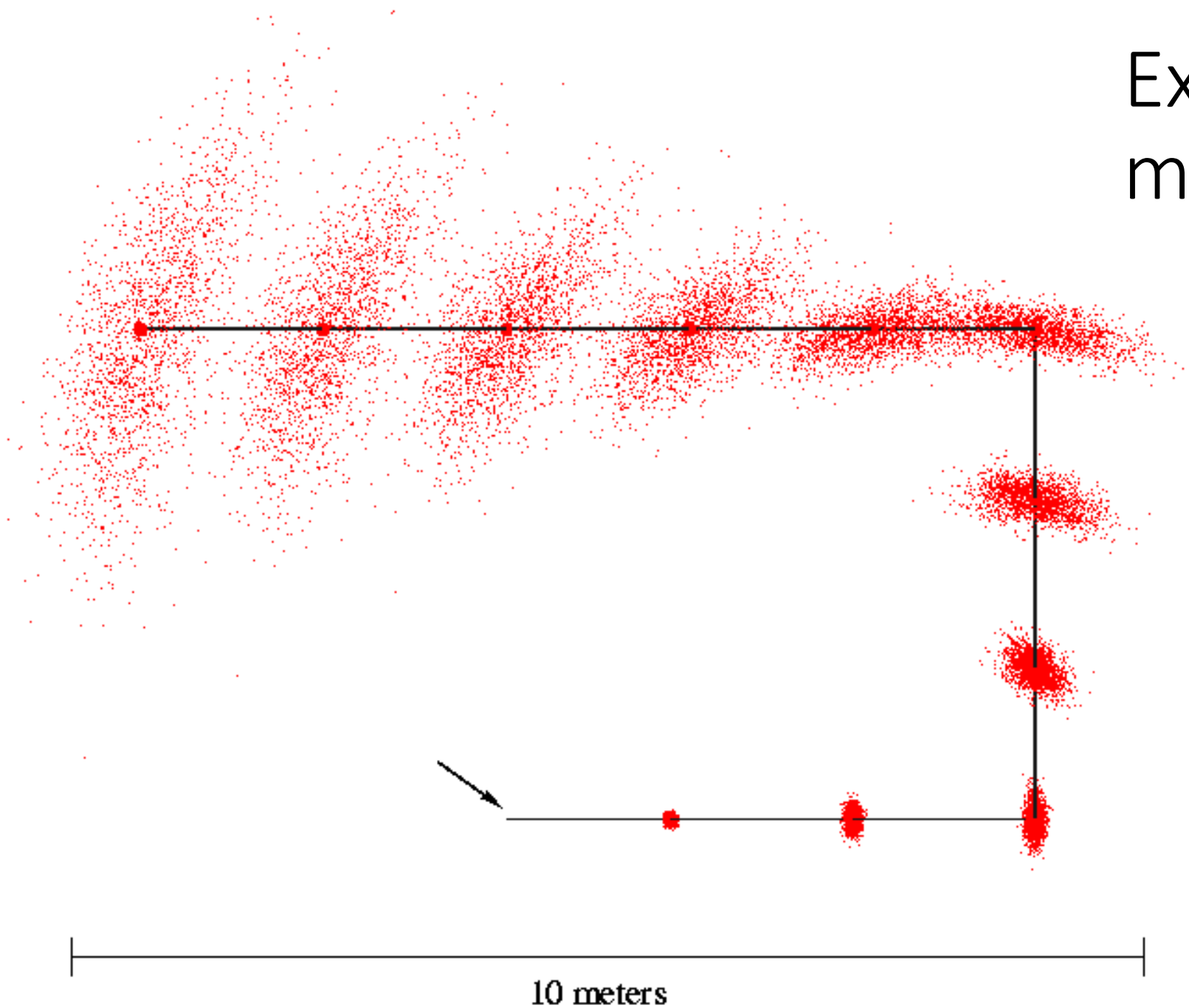
- a) Act  $p(x_2|x_1, u_1)$
- b) Sense  $p(z_2|x_2)$
- c) Sample from  $p(u_2)$

## 3. Slice 3:

- a) ...



# Example: motion model only



- The infamous “banana density”
- Happens because we also sample heading  $\theta$
- Clearly non-Gaussian!

# Summary

- Continuous Densities
- Gaussian Densities
- Bayes Nets & Mixture Models
- Cont. Measurement Models
- Cont. Motion Models
- Simulating Cont. Bayes Nets