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# **Boundary Detection Using Actuated Sensor Networks**

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**Introduction:** Control law for Gradient descent

#### **Sensor-based Planning**

Generalized Voronoi Graphs (Choset et. al.) have been used to

- perform sensor-based navigation of robots (fig 1)
- Provably convergent in n-dimensions

# Control Law has been proposed for such gradient descent

 $\dot{x} = \alpha \operatorname{Null}(\nabla G(x)) + \beta (\nabla G(x))^{\dagger} G(x)$ 

where  $(\nabla G(x))^{\dagger} = (\nabla G(x))^T (\nabla G(x) (\nabla G(x))^T)^{-1}$ 

Alpha, beta  $\rightarrow$  scalar gains  $G(x) = \begin{bmatrix} d_1(x) - d_2(x) \end{bmatrix}$  and  $\nabla G(x) = \left[ (\nabla d_1(x) - \nabla d_2(x))^T \right]$ 



Voronoi line using the control law

Fig 2 Gradient formation in the case of an actuated sensor node

**Adaptation to Sensor Networks** 

Control law modified to suit randomly deployed sensor networks G(x) = (sensor reading at the mobile node) - (threshold reading)defining the contour to be traced)

and  $\nabla G(x) = (\nabla d_1(x) - \nabla d_2(x))$ where

 $d_1 \rightarrow$  Unit vector in the direction of the steepest gradient towards the contour  $d_2 \rightarrow$  Unit vector in the direction of the steepest gradient away from the

contour

Each iteration performs one query of sensor readings on all neighboring sensor nodes to compute  $d_1$  and  $d_2$ (Fig 2)



Fig 3,4: Two cases of Boundary Detection using Actuated sensor node and the adapted version of the control law

## **Proposed Solution:** Adapt control law to perform boundary detection using actuated nodes

### **Assumptions**

· Each node is perfectly localized

• The sensed phenomenon is assumed to monotonically degrade in a certain predefined fashion

• Algorithm not drastically affected by type of gradients

### **Summary of results**

• Algorithm uses only local sensor information

•Saturation: Boundary detection percentage is above 80% for networks of degree > 6

• Optimality: Traversed path is within 10-20% of optimal path for networks with degree > 6

· Algorithm minimally affected by type of degradation in the gradient







Fig 3: Optimality of path taken by the actuated sensor node