

# Rise Aware Travelling Salesman Problem

**Abstract—**

**Index Terms—**Risk, Travelling Salesman

## I. INTRODUCTION

This document is a model and instructions for L<sup>A</sup>T<sub>E</sub>X. Please observe the conference page limits.

## II. PROBLEM 1: MAXIMISING THE CONDITIONAL VALUE AT RISK FOR A STOCHASTIC INFORMATION MAP

Consider a graph  $G(V, E)$ . A cycle is defined as a path that starts and ends at the same vertex. We define a Hamiltonian cycle  $C$  on this graph as a cycle that includes all vertices, passing through each vertex once. We call a function  $f$  submodular if:

$$f(S) + f(T) \geq f(S \cup T) + f(S \cap T) \quad (1)$$

The submodular function gives diminishing modular returns. Let us now define  $S$  as the set of information  $I(C)$  gained while travelling along  $C$ , the Hamiltonian cycle. We define  $f(S, y)$  as a sumodular function with  $y$  as the random variable associated with variance in the information gained. We can see that  $f(S, y)$  is a submodular function as the average information gained by travelling across edges must always be lesser than the average information gained from the edges seperately, that is

$$f(S \cup e) \leq f(S) + f(e) \quad (2)$$

We define the Value at Risk (VaR) as:

$$VaR_\alpha = \min_{\tau \in R} Pr[f(S, y) \leq \tau] \geq \alpha \quad (3)$$

and the Conditional Value at Risk (CVaR) as:

$$E_y[f(S, y) | f(S, y) \leq VaR_\alpha(S)] \quad (4)$$

While optimizing the CVaR, we instead optimize the auxiliary function  $H(S, \tau)$  defined as:

$$H(S, \tau) = \tau - \frac{1}{\alpha} E[(\tau - f(S, y))^+] \quad (5)$$

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### Algorithm 1 Risk TSP - 1

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**Input:**  $G(V, E), \tau_{max}, \alpha$ , information map  $\mathbf{M}$

**Output:**  $H$

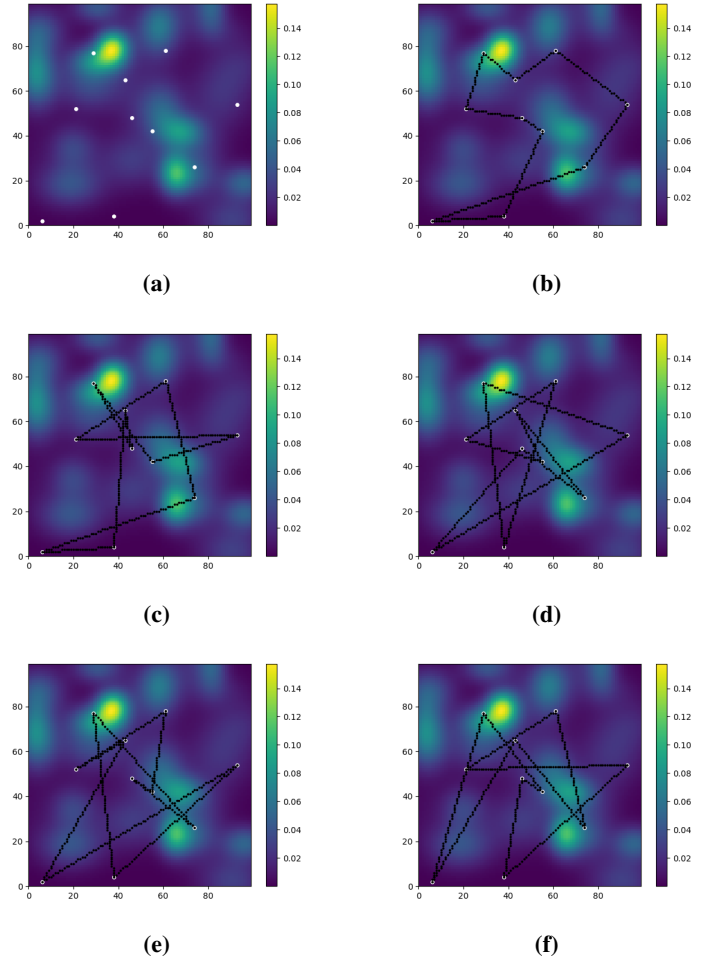
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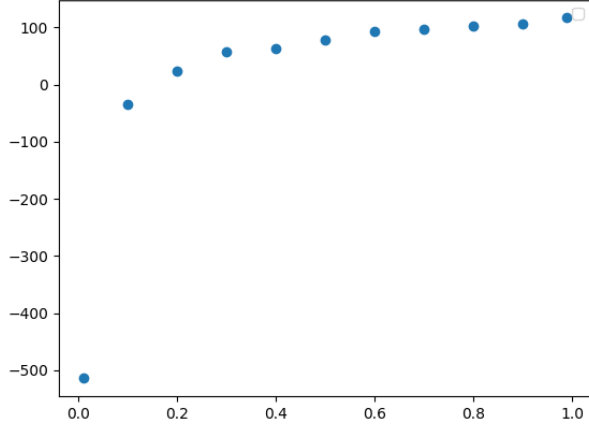
1: while  $\tau \leq \tau_{max}$  do
2:    $H_{all} = \emptyset$ 
3:    $f_{all} = \emptyset$ 
4:    $C = \emptyset$ 
5:    $H = 0$ 
6:    $f = 0$ 
7:   for  $e \in E$  do
8:     for 100 samples do
9:        $H((S \cup e), y) = \tau - \frac{1}{\alpha} E[(\tau - f(S, y))^+]$ 
10:       $f_{all}.append(avg(f))$ 
11:       $H_{all}.append(avg(H))$ 
12:    $H = max(H_{all})$ 
13:    $f = max(f_{all})$ 
14:    $C.append(e_{max})$ 

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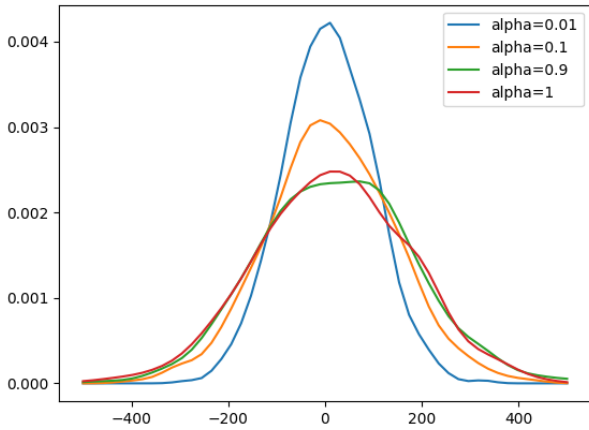
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**Fig. 1**



(a)



(b)

**Fig. 2**

### III. NEXT TO DO

Point discussed

- Implementation of stochastic risk for edges in graph  $G(V, E)$ , based on the Travelling Salesman formulation. We want to generate a Hamiltonian path with maximum reward given the expected risk along each edge (CVaR), and the operator threshold ( $\alpha$ ).
- Problem:  $\max \text{CVaR}_\alpha [w(S)], s \in H$
- Each edge reward is a random variable, e.g.,  $r_{eij} \sim N(10, 2)$
- Solution: CVaR minimization with greedy TSP as sub-routine
- Adding to the above algorithm costs, defined by constraints such as distance/fuel/time. These costs are deter-

ministic.

- Altering the costs to be expectations of stochastic values, and improving the above algorithm

### IV. CONCLUSION AND FUTURE WORK