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SR Number: 22205

UG BTech 3<sup>rd</sup> year

E0 259 Data Analytics

Quiz: 05/10/2024, 4:00 – 5:00 pm

Module: Mars Orbit

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|----|----|----|
| Q1 | Q2 | T  |
| 5  | 3  | 80 |
|    | 5  |    |

Instructions:

1. Please write your name and SR number on each page of the answer script now.
2. The quiz is in lieu of Assignment 2. There are two questions. Each carries five marks.
3. Each question is on a separate sheet of paper. Restrict your response to just that paper (front and back). You can use extra paper if you wish, don't attach them.
4. Please do not approach the TAs or the proctors for any clarifications. If there is something ambiguous, please state your assumptions and proceed to provide your response.
5. The duration of the test is 1 hr. No extra time will be given.

**Question 1:** Let  $lo_1, lo_2, lo_3, \dots, lo_k$  be the geocentric longitudes of Mars observed at times of  $k$  distinct oppositions that occur at times  $t_1, t_2, t_3, \dots, t_k$ , where the longitudes are in radians and times are in minutes, from some base year far enough back. Assume an equant at position  $(x, y)$  and the sun at  $(0, 0)$ . Assume Mars' orbit is a circle centred at the equant.

- a) What other unknown parameters do you need to fully define Mars' orbit?
- b) Write a pseudocode to derive the oppositions discrepancy at each of the  $k$  oppositions assuming the above orbit.

You can assume you have the following readymade functions; write any others you need:

- (i)  $Line(p_1, p_2)$ : Returns a line passing through two given points  $p_1, p_2$
- (ii)  $Point(l_1, l_2)$ : Returns the intersection point of two given lines  $l_1, l_2$
- (iii)  $Angle(l_1, l_2, p)$ : Returns the angle between lines  $l_1, l_2$ , both passing through  $p$
- (iv)  $Circle(l, p, r)$ : Returns the points of intersection of line  $l$  with a circle with centre  $p$  and radius  $r$ .

- 2 -  
a) To define mars' orbit, we need the radius of the orbit ( $r$ ) and the angular velocity of mars around the equant ( $\omega$ ). Additionally, to find mars' position at a given time, we would need to know its longitude ( $z$ ) at time  $t_1$  wrt the equant (assuming longitude taken from Sun-Aries line)

b) The opposition discrepancy is defined as the angle between the lines connecting sun to mars' observed & predicted position on the circle centred at the centre.



sun-aries reference

# Start with the longitudes observed from sun

$$\text{long-sun} = [l_{01}, l_{02}, \dots, l_{0k}]$$

# Assuming we have  $z$  <sup>initial long.</sup> &  $s$  <sup>angular speed</sup> fixed, get longitudes from equant.

$$\text{long-equant} = [z, z + (t_2 - t_1)s, z + (t_3 - t_1)s, \dots, z + (t_k - t_1)s] \% 2\pi$$

since its an angle  $\leq 2\pi$

# We have sun at (0,0), equant & centre at (x,y)

# Define a function line-fromangle

def line-fromangle (p<sub>1</sub>, a):

returns line passing through p<sub>1</sub> with angle a, from x-axis

# Now we need to calculate angle of equant's predicted mars from sun.

for i in range(k):

$$l_1 = \text{line-fromangle}((x,y), \text{long-equant}[i]) \quad \checkmark$$

$$p_1 = \text{circle}(l_1, (x,y), r) \quad \# \text{ assuming we fixed } r.$$

$$l_2 = \text{line}(p_1, (0,0)) \quad \# \text{ line from sun to equant pred}$$

$$l_3 = \text{line-fromangle}((0,0), \text{long-sun}[i]) \quad \# \text{ observed line from sun}$$

$$\delta_i = \text{angle}(l_2, l_3, (0,0)) \quad \# \text{ discrepancy}$$

Thus we get discrepancy  $\delta_i$  for all k times.

$$-9.5 + 0.5 = 3'$$

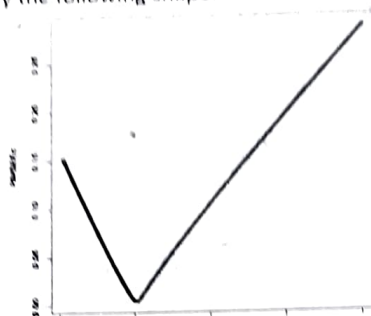
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Question 2: You are trying to fit an ellipse with focus at (0,0) to a few given points on a plane. The equation of the ellipse is given by the following in polar coordinates

$$r = l / (1 + ecc \cos(\theta))$$

where there are two parameters,  $l$  and  $ecc$ . Suppose  $l$  can vary from 10 to 20 and  $ecc$  can vary from 0.0 to 0.4. For any given value of  $l$  in the above range, the dependence of the root-mean-squared-error of the fit on  $ecc$  is given by the following shape.



(5)

Describe in pseudocode an algorithm to find a close to optimum value for  $l$  and  $ecc$ , with an eye on efficiency. You can assume you have access to a function  $RMSE(l, ecc)$  that, given  $l$  and  $ecc$  will give you the RMSE value of the fit.

# From the graph, it appears that there is only one optimum value of  $ecc$  for a given  $l$ . Also, the graph has a minima at the optimum point, & monotonic growth on ~~the~~<sup>both</sup> sides of minima.

# Thus, to find the best  $ecc$  for a given  $l$ , we need to perform gradient descent. However, since we only have one variable here, we can simply use a binary search algorithm.

def best\_ecc( $l$ , tolerance):

start = 0.0

end = 0.4

while  $|start - end| > tolerance$ :

$e = (start + end) / 2$

if  $RMSE(l, e) > RMSE(l, e + 0.00001)$ :

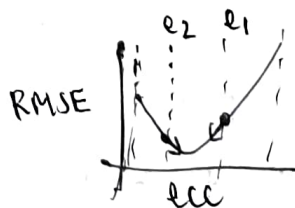
start =  $e$

else

end =  $e$

return  $(start + end) / 2$

# return final best ecc for  $l$



# RMSE of  $e$  & point right ahead of  $e$

# best-ec will return ~~the~~ ~~best~~ an ecc very close to the actual minima (within small tolerance). Since it is a binary search, it will converge fast.

# Now that we have best-ec for an  $l$ , we calculate it for all  $l$ 's in the given range.

```
best = []  
for l in range(10, 20, 0.01):  
    ecc = best-ec(l, tolerance)  
    best.append(RMSE(RMSE(l, ecc), l, ecc))
```

# 10 to 20 with step size 0.01  
Total 1000 values  
best ecc will give best RMSE for any  $l$ ?

# The above loop gives us best RMSE for all 1000 values of  $l$ . For best fit, we need RMSE to be the lowest. So we take top 1% RMSE and do a finer search on them.

```
sort(best) # sort in ascending order based on RMSE  
new = best[0:10] # Here new gets the corresponding l values for the top 1% RMSE
```

```
final = []  
for l1 in new:
```

```
    for l2 in range(l1-0.005, l1+0.005, 0.0001): # 100 values
```

```
        ecc = best-ec(l2, tolerance)
```

```
        final.append(RMSE(l1, ecc), l1, ecc)
```

# Now final has the best possible values of RMSE.

# select best RMSE from it.

```
sort(final) # sort in ascending order based on RMSE
```

```
best_fit = final[0] # lowest RMSE pair
```

Thus we get optimum value of  $l$ , ecc



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Note that we can make this more accurate by increasing the ~~to~~ ranges (ie 10000 values instead of 1000) for initial search.

For the final search, we can also choose the ranges to be a percentage interval (ie 99% to 101%) & shrink the range based on binary search upto some tolerance.