

# ⊕ TEXA - MATHEMATICAL REPRESENTATION :

## I -> TEST SERIES :

1 -> A test comprises of an Human Intelligence & Artificial Intelligence showcased by a Human Interrogator & the computer respectively.

$$\text{Let } f(m) = t_1 + t_2 + t_3 + t_4 + \dots + t_{n-1} + t_n$$

where  $m \rightarrow n^{\text{th}}$  transaction vector  $\phi$   $m = n/2$  &  $m \geq 1$   
 $n \geq 2$

Human & A.I correspondence is alternative in nature.

$$\text{i.e., } f(m) = t_1 + t_2 + t_3 + \dots$$

$$= A_0 + H_1 + A_2 + H_3 + \dots$$

where,

$A_0 + A_2 + A_4 + A_6 + \dots$  is the Even Series representing

transactions from an A.I :

$$\therefore T_n = 2(n-1) \Rightarrow A_{2(n-1)}$$

where,

$H_1 + H_3 + H_5 + H_7 + \dots$  is the Odd Series representing

transactions from an Human Interrogator :

$$\therefore T_n = 2n-1 \Rightarrow H_{2n-1}$$

( $\forall n \in \mathbb{N}, n \geq 1$ ),  $n$  is the Quantum Variable of the Transaction Series in the Test Instance. Hence,  $A_{2(n-1)}$  &  $H_{2n-1}$  are Quantum Scores.

## II. > VALUE DISTRIBUTION OF QUANTUM SCORES :

1. > Based on Game theory: A.I wins : +1 & Human loses : -1.  
A.I loses : -1 & Human wins : +1.

We have Quantum Scores :

$$\underline{A_{2(n-1)}} \quad \& \quad \underline{H_{2n-1}} \quad ( ; \forall n \geq 1 )$$

each contributing to the transaction series  $f(m)$ .

2. > Scheme of Value Distribution is as follows :

$$i. > \underline{A_{2(n-1)}} : = \begin{cases} 0 & ; \text{ if AI fails in deceiving human interrogator} \\ 1 & ; \text{ if AI successfully deceives human interrogator} \end{cases}$$

$$ii. > \underline{H_{2n-1}} : = \begin{cases} 1 & ; \text{ if Human interrogator recognizes AI} \\ 0 & ; \text{ if Human interrogator is deceived by the AI} \end{cases}$$

$m=3,$

3. > Eg : For  $n = 1, 2, 3, \dots, 6$ , let Quantum Scores for  $A_{2(n-1)}$  be 1, 1, 0, ...  
 $\therefore$  We get the transaction series :  
 $f(3) =$ 

1	0	1	0	0	1	.	.
$A_0$	$H_1$	$A_2$	$H_3$	$A_4$	$H_5$	$A_6$	$H_7$

for  $H_{2n-1}$  be 0, 0, 1, ...

### III → MEAN TEST SCORE - COMPUTATION :

#### 1) ARTIFICIAL INTELLIGENCE :

$$\begin{array}{c} M \cdot T \cdot S \\ \text{(MEAN TEST SCORE)} \end{array} = \frac{\sum_{n=1}^N A_{2(n-1)}}{N} \quad ( ; \forall N \geq 2 \text{ transactions} )$$

#### 2) HUMAN INTELLIGENCE :

$$\begin{array}{c} M \cdot T \cdot S \\ \text{(MEAN TEST SCORE)} \end{array} = \frac{\sum_{n=1}^N H_{2n-1}}{N} \quad ( ; \forall N \geq 2 \text{ transactions} )$$

#### 3) M-T-S MATRIX :

$$\begin{array}{c} \begin{array}{c} H \cdot I_1 \quad H \cdot I_2 \quad H \cdot I_3 \quad \dots \quad H \cdot I_j \\ \downarrow \end{array} \\ \begin{array}{c} A \cdot I_1 \\ A \cdot I_2 \\ A \cdot I_3 \\ \vdots \\ A \cdot I_i \end{array} \left[ \begin{array}{cccc} (, ) & (, ) & (, ) & \dots (, ) \\ & & & \dots (, ) \\ & & & & \vdots \\ & & & & \vdots \\ (, ) & & & \dots (, ) \end{array} \right]$$

## IV → CATEGORIZATION :

1. An A.I can exhibit its intelligence in multiple directions at multiple levels.
2. Categorization & Category scores are critically important in understanding the behavior of A.I in multiple environments.

### 3. → MULTI - CATEGORY CLASSIFICATION LOGIC :

PROBLEM : How to classify an A.I /nominate it for multiple categories, all under one test (1 or more) ?

SOLUTION :- By measuring the performance of A.I in handling the questions (ordered in many ways).

- "Slab" is a set of questions pertaining to a category /domain.
- Slab consists of questions in an ordered /shuffled manner.
- Slabs can also be shuffled in the test instance.

$$- \text{S.P.F}_{\text{SLAB}_i} = \frac{(\text{RESPONSE FACTOR})_{\text{SLAB}_i}}{(\text{ERROR FACTOR})_{\text{SLAB}_i}}$$

$$\left( \begin{array}{c} \text{SLAB} \\ \text{PERFORMANCE} \\ \text{FACTOR} \end{array} \right)$$

$$- (\text{RESPONSE FACTOR})_{\text{SLAB}_i} = \frac{\sum_1^{\text{MEAN SLAB SIZE}} \text{No. OF Q's ANSWERED (i.e., } A_{2(n-1)}) \left[ \forall n \in \text{SLAB}_i \right]}{\text{TOTAL No. OF Q's IN THE SLAB (SLAB SIZE)}}$$

$$- (\text{ERROR FACTOR})_{\text{SLAB}_i} = \frac{\sum_1^{\text{MEAN SLAB SIZE}} \text{No. OF Q's FAILED (i.e., } A_{2(n-1)}) \left[ \forall n \in \text{SLAB}_i \right]}{\text{IN ALL TEST INSTANCES}}$$

where, SLAB SIZE =  $n(\text{SLAB}_i)$  & MEAN SLAB SIZE =  $\frac{\sum_{i=1}^u n(\text{SLAB}_i)}{u}$ , where  $u \rightarrow$  No. of slab; occurrence in 'v' no. of test instance.

#### IV - CONT'D :-

40) Logic Result :

LOGIC   RESULT :

- If  $S.P.F. / E_i =$   $\begin{cases} < 0 : \text{Result} = -1 \quad ; \text{i.e., AI non-performant} \\ > 0 : \text{Result} = S.P.F. \% \quad ; \text{i.e., AI performant to the category.} \end{cases}$

5. CATEGORY MATRIX:

$$A \cdot I_1 \begin{bmatrix} C_1 & C_2 & C_3 & \dots & C_y \\ \forall x \in Z^+, & A \cdot I_x \notin C_y \text{ iff } E_i = -1 & & & \\ A \cdot I_2 & & \text{OR} & & \\ \forall x \in Z^+, & & & & \\ A \cdot I_3 & & & & \\ \vdots & & & & \\ A \cdot I_x \in C_y \text{ iff } E_i \neq -1 & & & & \\ \text{where } i = (x \times y) \text{ Row Column Index.} & & & & \\ \vdots & & & & \\ A \cdot I_x & & & & E_{xy} \end{bmatrix}$$

where,

A-1  $\rightarrow$  Artificial Intelligence under Turing Test 4

$C \rightarrow$  Category / Domain to which A.I is tested / not tested.

## V → RANKING :

### 1. > GENERIC RANKING :

$$\begin{aligned} \text{Generic Performance (per AI)} &= \frac{\sum_{i=1}^v (M \cdot T \cdot S) \text{AI}_i / v}{\sum_{i=1}^v (M \cdot T \cdot S) \text{HI}_i / v} = \frac{\sum_{i=1}^v (M \cdot T \cdot S) \text{AI}_i}{\sum_{i=1}^v (M \cdot T \cdot S) \text{HI}_i} \end{aligned}$$

For  $x$  AIs, rankings are ordered based on the Performance.

### 2. > CATEGORY RANKING :

$$\text{Category Performance} = \text{S.P.F}$$

For  $x$  AIs, category rankings are ordered based on SPF Values.



## VI → GRADINGS : (APPLICATION OF TERA THEORY)

1. Grading A.I.s will be quintessential in limiting their behaviors in certain environments that are tangibly insecure - physically or mentally.
2. The S.P.F parameters aids in understanding the strength of the A.I (Responsiveness) in a particular domain.
3. Some objects may be common to 2 or more ( $\geq 2$ ) domains. This must not compromise the security of the environment since multiple A.I.s/profiles may be deployed onto an instance/robot.

4. 'G' :

- A Factor to specify the developer to customize boundaries, thereby providing flexible logic & automation opportunities.
- Let an A.I exhibit its intelligence under 2 categories  $C_p$  &  $C_q$ , where  $C_p$  &  $C_q$  correspond to  $Slab_p$  &  $Slab_q$  with an array of questions in a user-defined order.
- Now,  $Slab_p$  &  $Slab_q$  may have common objects such as "knife" or other dangerous objects.
- Corresponding  $A_{2(n-1)}$  quantum values of such questions are collected from  $p$  &  $q$ .
- Collected  $A_{2(n-1)}$  values of corresponding objects in the questions are operated under ~~AND~~ <sup>AND</sup> gate:
- $A_{2(i-1)}$  from  $Slab_p$  &  $A_{2(j-1)}$  from  $Slab_q$  may have common object "knife".

$$\therefore G = A_{2(i-1)} \left( \begin{matrix} \text{AND} \\ \text{AND} \end{matrix} \right) A_{2(j-1)} \quad \left| \text{If } G = \begin{cases} 0 & ; \text{ Safe to use, no logic overlapping} \\ 1 & ; \text{ Unsafe since overlapping logic.} \end{cases} \right.$$