

# Multiprocessing Suited Pace Size Proficiency for Ciphering First Order ODEs

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**Abstract**-Appraising computed error in forecasting-adjusting system is all essential for purposeful acquiring suited pace size. Diverse schemata for controlling/estimating error bank on forecasting-adjusting system. This study examines multiprocessing suited pace size (adaptive) proficiency for ciphering first order ordinary differential equations (ODEs). This involves compounding Newton's back difference interpolating multinomial with numeral consolidation method. This is valued at more or less preferred grid points to invent multiprocessing forecasting-adjusting system. Moreover, process progresses to produce main local truncation error (MLTE) of multiprocessing forecasting-adjusting system after showing degree of the system. Numeral resolutions manifest effectiveness of varying pace size in working out first order ODEs. Accomplished resolution rendered is aided using mathematical software program. Mathematical resolutions show-case adaptive proficiency is effectual and function better than subsisting systems with respect to the maximal computed errors in the least time-tested tolerance bounds.

**Index Terms**- Adaptive proficiency, forecasting-adjusting system, tolerance bound, maximal computed errors, main local truncation error.

## 1. INTRODUCTION

Pace length alternative operation formulates close together the main local truncation error that does not postulate calculation of higher derived function. Proficiencies defined as adaptive, requires modification of amount and positioning of grids utilized in estimating primary local truncation error tween specified bound. Nevertheless, approximate close relationship upholds inside problem of

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calculating value of explicit integral and ciphering resolution to initial value-problem. Adaptive proficiencies are no surprising for computing resolutions to ODEs and these proficiencies are not exclusively effective but integrate maximal control error. See [1], [12].

Forecasting-adjusting system is a substituted method for meaningful estimating maximal control error thru worthy pace length. Paper concerned computing maximal control error of first order ODEs of the type [11]

$$g' = \bar{z}(t, g), \quad c \leq t \leq d, \quad g(t_0) = g_0. \quad (1)$$

Computational resolution of (1.1) in general is

given as

$$\sum_{i=0}^r \alpha_i g_{n+i} = h^* \sum_{i=0}^r \beta_i \bar{z}_{n+i}, \quad (2)$$

where pace-size is  $h^*$ ,  $\alpha_r = 1$ ,  $\alpha_i$ ,  $i = 0, 1, \dots, r$ ,  $\beta_r$  are unknown quantity specifying such pattern of degree  $r$ . See [4].

Assuming  $\bar{z} \in R$  to a sufficient degree possessing differential coefficient on  $t \in [c, d]$  and meets global

Lipchitz consideration, i.e., there is invariable  $\bar{L} \geq 0 \ni |\bar{z}(t, \check{g}) - \bar{z}(t, \bar{g})| \leq \bar{L}|\check{g} - \bar{g}|, \quad \forall \check{g}, \bar{g} \in R. (3)$

Underneath this presumptuousness, (1) insure existence and singularity fixed on  $t \in [c, d]$  likewise looked at fulfilling Weierstrass theorem. See for example [15], [18], [26] for details.

Where  $c$  and  $d$  are bounded,  $g' = [g'_1, g'_2, \dots, g'_n]^T$ ,  $\check{g} = [\check{g}_1, \check{g}_2, \dots, \check{g}_n]$  and  $\bar{z} = [\bar{z}_1, \bar{z}_2, \dots, \bar{z}_n]$ , spring-up in real-life application problems in scientific discipline and engineering science like mobile dynamical and motion of rocket as presented [5]-[6], [22].

The intention requires adopting adaptive proficiency to invent varying pace-size of multiprocessing forecasting-adjusting system. Technics of continuing in varying pace-size forecasting-adjusting system commenced with Milne's device. Look at [11]-[12], [18]-[19] for more particulars. Varying pace length describe Milne's device valuation of ODEs, thus rest on diverse elements as considered in [23]-[24], [27]. Scholarly person suggested adaptive proficiencies in elementary class of Adams typecast. Search [2]-[3], [7], [9], [16], [23]-[25]. Nevertheless, bookmen like [2]-[3], [5]-[7], [9], [16] possesses many disfavours which include inability to design/vary a suitable pace size, check tolerance bound and yield better maximal calculated errors. Gear's method acknowledged backward differentiation formula (BDF) for stiff problems. See [14], [20]-[22], [27]-[28].

Motivation of research paper propose adaptive proficiencies of forecasting-adjusting system for estimating nonstiff and mildly stiff ODEs implemented in  $P_f^\sigma(E^\sigma C_A^\sigma)^j$  or  $P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$  style.

Proficiency possess some vantages alike designing/varying suited pace-size, specifying tolerance bound and controlling computed error besides dealing with shortcomings stated above. Look into [2]-[3], [7], [9], [16] for more items.

This paper structure is studied as complies: in Subdivision 2 Substance and Techniques. Subdivision 3 Numeric Examples. Subdivision 4 Effect and Discourse. Subdivision 5 Decision as mentioned [4], [24].

## II. SUBSTANCE AND TECHNIQUES

This subdivision deploys Newton's back difference formula to devise forecasting-adjusting system.

Presuppose  $\bar{z}(t)$  possess uninterrupted  $\bar{k}th$  differential coefficient,  $t_j = t_0 + j\bar{h}$ ,  $\bar{z}_j = \bar{z}(t_j)$ , and back differences represent

$$\nabla^{\bar{q}+1}\bar{z}_j = \nabla^{\bar{q}}\bar{z}_j - \nabla^{\bar{q}}\bar{z}_{j-1}$$

where  $\nabla^0\bar{z}_j = \bar{z}_j$ , then

$$z(t) = z_j + \left(\frac{t-t_j}{h}\right)\nabla z_j + (t-t_j)(t-t_j - \frac{\nabla^2 z_j}{2!h^2} + \dots + (t-t_j) \dots (t-t_{j-k+2}) \frac{\nabla^{k-1} z_j}{(k-1)!h^{k-1}} + (t-t_j) \dots (t-t_{j-k+1}) \frac{z^{(k)}(\vartheta)}{k!}, \quad (4)$$

where  $z^{(k)}(\tau)$  is the  $\bar{k}th$  derived function of  $z$  assessed at some point interval possessing  $t, t_{j-k+1}, t_j$ . Presume  $x = \frac{(t-t_{n-1})}{h}$ ,  $j = n-1$ , (2.1) gives

$$z(t) = \binom{-x}{0} z_{n-1} - \binom{-x}{1} \nabla z_{n-1} + \dots + (-1)^{k-1} \binom{-x}{k-1} \nabla^{k-1} z_{n-1} + (-1)^k h^k \binom{-x}{k} z^{(k)}(\vartheta)$$

$$\binom{x}{k} = \frac{x(x-1)\dots(x-q+1)}{q!}, \binom{x}{0} = 0.$$

Replacing supra in

$$g(t_n) = g(t_{n-1}) + \int_{t_{n-1}}^{t_n} z(t)dt,$$

to obtain

$$g(t_n) = g(t_{n-1}) + \int_{t_{n-1}}^{t_n} \left[ \sum_{m=0}^{k-1} (-1)^j \binom{-x}{m} \nabla^m f_{n-1} + (-1)^k h^{*k} \binom{-x}{k} g^{(k+1)}(\vartheta) \right] dt$$

$$g(t_n) = g(t_{n-1}) + h^* \sum_{m=0}^{k-1} \gamma^m \nabla^m z_{n-1} + (-1)^k h^{*k} \int_0^1 \binom{-x}{k} g^{(k+1)}(\vartheta) dx \quad (5)$$

Every time conclusive term supra is dismissed, the left over will be called  $\rho$  - stride Adams-Bashforth formula

$$g_n = g_{n-1} + \bar{h} \sum_{m=0}^{k-1} \gamma^m \nabla^m z_{n-1}. \quad (6)$$

Back differences show with regards to valuates at extend of proceeding degrees

$$\nabla^q z_{n-1} = \sum_{i=0}^q (-1)^i \binom{q}{i} z_{n-1-i}$$

(6) is penned

$$g_n = g_{n-1} + \bar{h} \sum_{m=0}^{k-1} \beta_{ki-i} z_{n-i} \quad (7)$$

Proceeding with (2.4) produce multiprocessing  $\rho$ -step Adams-Bashforth Formula [8].

Likewise, adjusting multiple stride systems-Adams-Moulton system is deduced setting  $j = n$  into (4) and subbing

$$g(t_n) = g(t_{n-1}) + \int_{t_{n-1}}^{t_n} z(t)dt,$$

yields

$$g(t_n) = g(t_{n-1}) + \int_{t_{n-1}}^{t_n} \left[ \sum_{m=0}^{k-1} (-1)^m \binom{-x+1}{m} \nabla^m z_n + (-1)^k h^{*k} \binom{-x+1}{k} g^{(k+1)}(\vartheta) \right] dx$$

Disregarding computed error term, yields method

$$g_n = g_{n-1} + h^* \sum_{m=0}^{k-1} \gamma_m^* \nabla^m z_n \quad (8)$$

Subbing  $\nabla^m z_n$  in terms  $z_n, z_{n-1}, z_{n-2}, \dots$ , generates

$$g_n = g_{n-1} + \bar{h} \sum_{i=0}^{k-1} \beta_{ki}^* z_{n-i}. \quad (9)$$

In continuance, (9), multiprocessing  $\rho$  -1-stride Adams-Moulton system is generated. See [13].

**Order of accuracy-** Adopting [4], [18], consociated linear multiple strides system (2) and difference mathematical expression

$$\hat{L}[g(t); \bar{h}] = \sum_{i=0}^k \alpha_i g_{n+i} - \bar{h} \beta_k z_{n+i}. \quad (10)$$

Accepting  $g(t)$  is sufficiently and endlessly possessing a derivative on time limit  $[c, d]$ . This implies that  $g(t)$  owns numerous derived function demanded, composing terminus in (10) as Taylor series expression of  $g(t_{n+i})$  and  $z(t_{n+i}) \equiv g'(t_{n+i})$   $g(t_{n+i}) = \sum_{k=0}^{\infty} \frac{(i\bar{h})}{k!} g^{(k)}(t_n)$  and  $g'(t_{n+i}) = \sum_{k=0}^{\infty} \frac{(i\bar{h})}{k!} g^{(k+1)}(t_n)$  (11)

Standing in (10), (11) into (2), succeeding formulation obtains  $\hat{L}[g(t); \bar{h}] = C_0 g(t) + C_1 \bar{h} g'(t) + C_2 \bar{h}^2 g''(t) + \dots + C_p \bar{h}^p g^{(p)}(t) + C_{p+1} \bar{h}^{p+1} g^{(p+1)}(t) \dots$  (12)

Concurring [4], [18], adaptive proficiencies (2) gives degree  $p$ , whenever  $C_{p+1}, p = 0, 1, 2, \dots, i = 1, 2, \dots, m$ , establish following:

$C_{p+1}, p = 0, 1, 2, \dots, i = 1, 2, \dots, m$ , yielded adopting:

$$C_0 = \alpha_0 + \alpha_1 + \alpha_2 + \dots + \alpha_k,$$

$$C_1 = \alpha_0 + 2\alpha_1 + \alpha_2 + \dots + k\alpha_k,$$

$$C_2 = \frac{1}{2!} (\alpha_0 + \alpha_1 + \alpha_2 + \dots + \alpha_k) - (\beta_0 + \beta_1 + \beta_2 + \dots + \beta_k),$$

$$C_q = \frac{1}{q!} (\alpha_1 + 2^q \alpha_2 + \dots + k^q \alpha_k) - \frac{1}{(q-1)!} (\beta_0 + 2^{q-1} \beta_1 + \dots + k^{q-1} \beta_k), \quad q = 2, 3, \dots$$

Thusly, (2) gives degree  $p \geq 1$  and computed error constant quantity  $C_{q+1} \neq 0$ .

Harmonizing [18], [19], system (2), degree  $p$  conditionally

$$\hat{L}[g(t); \bar{h}] = 0\bar{h}^{p+1}, \quad C_0 = C_1 = C_2 = \dots C_p = 0, \quad C_{p+1} \neq 0. \quad (13)$$

Therefore,  $C_{p+1}$  is computed error constant quantity and  $C_{p+1}\bar{h}^{p+1}g^{(p+1)}(t_n)$  main principal local truncation error at  $t_n$ .

**Stableness Property-** Properly stableness system is required in multiprocessing forecasting-adjusting system to anneal. This follows scripted multiprocessing system in establishing rectangle array of bounded difference. See [4], [17].

$$C^{(0)}G_j = C^{(1)}G_{j-1} + \bar{h}(D^{(0)}Z_m + D^{(1)}Z_{m-1}), \quad (14)$$

$$G_j = \begin{bmatrix} g_{n+1} \\ g_{n+2} \\ \vdots \\ g_{n+r} \end{bmatrix}, \quad G_{j-1} = \begin{bmatrix} g_{n-r+1} \\ g_{n-r+2} \\ \vdots \\ g_n \end{bmatrix},$$

$$Z_j = \begin{bmatrix} z_{n+1} \\ z_{n+2} \\ \vdots \\ z_{n+r} \end{bmatrix}, \quad Z_{j-1} = \begin{bmatrix} z_{n-r+1} \\ z_{n-r+2} \\ \vdots \\ z_n \end{bmatrix}.$$

Rectangular arrays  $C^{(0)}, C^{(1)}, (D^{(0)}, D^{(1)})$  determine  $r \times r$  rectangular arrays having actual substantial. Altho  $G_j, G_{j-1}, Z_j, Z_{j-1}$  gives variable quantity stated supra.

Accordingly [17]-[18], boundary locus system fix region of absolute stability of multiprocessing system is carried out to generate root system of absolute stability. Exchanging test par  $g' = -\tau g$  and  $\bar{h} = h^* \tau$

into multiprocessing (14), resulting to

$$\phi(r) = \det [r(C^{(0)} + D^{(0)}h^*\tau) - (C^{(1)} + D^{(1)}h^*\tau)] = 0. \quad (15)$$

Subbing  $\bar{h} = 0$  in (15), roots system generated  $\tau \leq 1$ . Hence, from definition [18]-[19], absolutely stability is gratified.

Moreover, from [15], edge region of absolute stability is incurred interchanging (2) within

$$\bar{h}(r) = \frac{\psi(r)}{\sigma(r)} \quad (16)$$

and  $r = e^{i\theta} = \cos \theta + i \sin \theta$  simplified unitedly and valuating (16) inside  $[0^0, 180^0]$ . Therefore, bounds of region of absolute stability dwells on actual axis.

**Practical Implementation of Adaptive Proficiencies**  
Concurring with [13], [15]-[16], practical implementation of  $P_f^\sigma(E^\sigma C_A^\sigma)^j$  or  $P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$  fashion gets all-important in forecasting and adjusting systems assuming both are of ilk degree. Since this precondition is of essence for stride-number of forecasting unit higher than adjusting

system. Consequently, fashion of  $P_f^\sigma(E^\sigma C_A^\sigma)^j$  or

$P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$  is took apart as adopts,  $j = 1, 2, \dots$  :

$$P_f^\sigma(E^\sigma C_A^\sigma)^j:$$

$$g_{n+m}^{[0]} + \sum_{i=0}^{m-1} \alpha_i \cdot g_{n+i}^{[j]} = h^* \sum_{i=0}^{m-1} \beta_i \cdot z_{n+i}^{[j-1]},$$

$$z_{n+m}^{[s]} \equiv z(x_{n+m}, g_{n+m}^{[s]}),$$

$$g_{n+m}^{[s+1]} + \sum_{i=0}^{m-1} \alpha_i g_{n+i}^{[j]} = h^* \alpha_m z_{n+m}^{[s]} + h^* \sum_{i=0}^{m-1} \beta_i z_{n+i}^{[j-1]}, \quad s = 0, 1, \dots, j-1 \quad (17)$$

$$P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$$

$$g_{n+m}^{[0]} + \sum_{i=0}^{m-1} \alpha_i \cdot g_{n+i}^{[j]} = h^* \sum_{i=0}^{m-1} \beta_i \cdot z_{n+i}^{[j]},$$

$$z_{n+m}^{[s]} \equiv z(x_{n+m}, g_{n+m}^{[s]}),$$

$$g_{n+m}^{[s+1]} + \sum_{i=0}^{m-1} \alpha_i g_{n+i}^{[j]} = h^* \beta_m z_{n+m}^{[s]} + h^* \sum_{i=0}^{m-1} \beta_i z_{n+i}^{[j]}, \quad s = 0, 1, \dots, j-1,$$

$$z_{n+m}^{[j]} \equiv z(x_{n+m}, g_{n+m}^{[j]})$$

Mentioning as  $j \rightarrow \infty$ , resolution of computing with alternatives supra fashion will gear towards adjusting to tolerance bound.

What is more, forecasting and correcting system centered on (2) is enforced.  $P_f^\sigma(E^\sigma C_A^\sigma)^j$  or  $P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$  is specified by (17),  $h^*$  is pace size which assumes that forecasting and adjusting system is of alike degree  $p^\sigma$ .

Incases,  $C_{p+1}^F, C_{p+1}^A$  yields workout error constant quantity of forecasting-adjusting system in given order. The complying consequence applies

**Proposition-** Presuppose forecasting system owns degree  $p^F$  and adjusting system gets degree  $p^A$  utilized in  $P_f^\sigma(E^\sigma C_A^\sigma)^j$  or  $P_f^\sigma(E^\sigma C_A^\sigma)^j E^\sigma$  style,  $p^F, p^A, m$  are whole numbers and  $p^F \geq 1, p^A \geq 1, m \geq 1$ . Thus, setting conditionally for

$p^F \geq p^A$  (or  $p^F < p^A$ ,  $m > p^A - p^F$ ) brings forth the forecasting-adjusting systems having ilk degree and ilk MLTE as adjusting system. Pre-conditionally for  $p^F < p^A$  and  $m = p^A - p^F$ , forecasting-adjusting system owns ilk degree as the adjusting with unlike MLTE.

Supposedly for  $p^F < p^A$  and  $m \leq p^A - p^F - 1$ , forecasting-adjusting system possessed alike degree equated to  $p^F + m$  but to a lesser extent than  $p^A$ .

In distinction from others, noting that forecasting system possesses degree  $p^F$  and adjusting system degree  $p^A$  as  $P_f^\sigma (E^\sigma C_A^\sigma)^j$  consequentially gives birth to degree  $p^\sigma$ . In addition,  $P_f^\sigma (E^\sigma C_A^\sigma)^j$  or  $P_f^\sigma (E^\sigma C_A^\sigma)^j E^\sigma$  system owns ilk degree and ilk MLTE as discoursed [15]-[16]. Mixing [9]-[10], [15]-[16], Milne's device put forward viability to calculate MLTE of forecasting and adjusting systems in absence of higher differential coefficients,  $g(t)$ .

Setting  $p^A = p^F$ ,  $p^F$  and  $p^A$  expresses degree of forecasting and adjusting systems with ilk order. Forthwith, degree  $p^\sigma$ , MLTE is spelt infra

$$C_{p+1}^F \bar{h}^{-p+1} g^{(p+1)}(t_n) = g(t_{n+j}) - \bar{W}_{n+m} + o(\bar{h}^{-p+2}), \quad (18)$$

$$C_{p+1}^A \bar{h}^{-p+1} g^{(p+1)}(t_n) = g(t_{n+m}) - C_{n+m} + o(\bar{h}^{-p+2}) \quad (19)$$

$W_{n+m}^F$  and  $C_{n+m}^A$  refers to forecasting and adjusting approximates established by degree  $p^\sigma$ ,  $C_{p+1}^F$  and  $C_{p+1}^A$  distinct of  $\bar{h}$ .

Premitting terminus of degree  $P^\sigma + 2$  supra and possibly make computational approximate of MLTE of the system as

$$C_{p+1}^A \bar{h}^{-p+1} g^{(p+1)}(t_n) = \frac{C_{p+1}^A}{C_{p+1}^F - C_{p+1}^A} |W_{n+m}^F - C_{n+m}^A| < \bar{\delta} \quad (19)$$

Mentioning  $C_{p+1}^F \neq C_{p+1}^A$  and  $W_{n+m}^F \neq C_{n+m}^A$ .

All the same, evaluating MLTE (19) establishes acceptance and rejection of the output or ingeminating the pace with a lesser pace size. The pace permissible is founded on a try out prescribed by (19) as seen [8]. (19) yields tolerance bound of Milne's estimate for adjusting to convergence.

### III. NUMERIC EXAMPLES

**Problem tested-** Three tested problems have been adopted and carry out utilizing ATVS-SBP-CM on two distinct convergence criteria 0.0001 and 0.000001 as viewed [28].

Tested problem 1  $g'(t) = -0.5g$ ,  $g(0) = 1$ ,  $0 \leq t \leq 20$ .

Exact Solution:  $g(x) = e^{-0.5t}$

Tested problem 2  $g'(t) = -g$ ,  $g(0) = 0$ ,  $0 \leq t \leq 20$ .

Exact Solution:  $g(t) = e^{-t}$

Test problem 3  $g'(t) = -300tg$ ,  $g(0) = 1$ ,  $0 \leq t \leq 20$ .

Exact Solution:  $g(t) = e^{-150t^2}$ .

### IV. EFFECTS AND DISCOURSE

In subsection, computational resolutions show execution of adaptive proficiencies for figuring first-order ODEs. Finish resolution yielded is found employing computational software package. The language used is enlisted under:

Table 1, Table 2 and Table 3 presents the numeric decisions of problems 1, 2, and 3 applying APVPSMF-AS in comparison to subsisting systems. Acronyms listed on table 1, table 2 and table 3 are enlisted infra.

**Table 1 of Problem 1**

Proficiency	Max <sub>computederror</sub>	T <sub>olb</sub>
GMBDF	1.81471e - 04	10 <sup>-4</sup>
BGMBDF	1.24969e - 04	
APVPSMF-AS	1.1823e - 04	10 <sup>-4</sup>
APVPSMF-AS	1.08184e - 04	
GMBDF	1.26832e - 06	10 <sup>-6</sup>
BGMBDF	1.24997e - 05	
APVPSMF-AS	1.24578e - 06	10 <sup>-6</sup>
APVPSMF-AS	1.16188e - 06	

**Table 2 of Problem 2**

Proficiency	Max <sub>computederror</sub>	T <sub>olb</sub>
GMBDF	2.45713e - 04	10 <sup>-4</sup>
BGMBDF	2.49875e - 04	
APVPSMF-AS	1.08184e - 04	10 <sup>-4</sup>
APVPSMF-AS	1.06312e - 04	
GMBDF	2.17301e - 06	10 <sup>-6</sup>
BGMBDF	2.49988e - 05	
APVPSMF-AS	1.24578e - 06	10 <sup>-6</sup>
APVPSMF-AS	1.16188e - 06	

**Table 3 of Problem 3**

Proficiency	Max <sub>computederror</sub>	T <sub>olb</sub>
GMBDF	6.66952e - 04	10 <sup>-4</sup>
BGMBDF	2.82904e - 04	
APVPSMF-AS	8.82506e - 05	10 <sup>-4</sup>
APVPSMF-AS	8.36745e - 05	
GMBDF	7.14465e - 07	10 <sup>-6</sup>
BGMBDF	1.91068e - 07	
APVPSMF-AS	1.08108e - 07	10 <sup>-6</sup>
APVPSMF-AS	1.04717e - 07	

APVPSMF-AS: adaptive proficiencies of suited pace size multiprocessing forecasting-adjusting system.

T<sub>olb</sub>: tolerance bound.

P<sub>rof</sub>: proficiency applied.

Max<sub>computederror</sub>: magnitude of maximal computed errors in APVPSMF-AS.

GMBDF: computed error in GMBDF (Generalized Multistep Adams and Backward Differentiation Formulae) for tested problems 1 and 2 as seen [28].

BGMBDF: computed error in BGMBDF

(Block Method for Generalized Multistep Adams and Backward Differentiation Formulae) for test problem 3 as seen [28].

## V. DECISION

Computational effects showed APVPSMF-AS is implemented utilizing tolerance bound. Tolerance bound settles acceptance or rejection of resolution. Computed effects similarly launch functioning APVPSMF-AS is remarked to bring forth improve maximal computed errors liken to BGMBDF and

GMBDF as cited [28]. Whence, deciding multiprocessing system invented having qualities for figuring out nonstiff and stiff ODEs as remarked supra.

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