Analytical Model for the Effects of Reservoir Formation Matrix on the Performance of Microbial Enhanced Oil Recovery

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ABSTRACT: The performance of any EOR technique puts into critical consideration the reservoir geology, lithology, reservoir uniformity, pay continuity and a wide range of other rock properties. This multi-mechanism, environmental friendly and inexpensive biotechnological approach to oil recovery (MEOR) must also put into consideration the formation characteristics for an effective and productive microbial injection project. This study tends to ascertain numerically the effects of formation matrix on the performance of MEOR by presenting a mathematical model accounting for microbial concentration distribution for various formation types. Results showed that for a homogenous reservoir, the average microbial concentration in the reservoir increased linearly with increasing days of microbial injection. Variation of porosity and permeability in heterogeneous reservoir yielded a non-linear relationship as a result of the varying rock properties when averaging method of permeability and porosity was adopted. A non-averaged permeability and porosity deduction for a heterogeneous formation will result in a more distorted concentration profile of the injected microbes. An irregularity in concentration distribution of microbes in these heterogeneous reservoirs is traceable to the distorted propagation and transportation of these injected microbes.

KEYWORDS: Formation, Heterogeneous, Homogeneous, Matrix, MEOR, Microbes

I. LITERATURE SURVEY

The rock properties and their variations from one location to another significantly influence MEOR performance and ultimately the oil recovery [6]. Fredrickson et al presented experimental results on pore-size constraints on the activity and survival of the subsurface bacteria using different grain size of the medium to high permeability [4], [7]. The microbes were investigated in fractured porous media using etched-glass micro-models [8], [8] also used non fracture models to compare the efficiency of MEOR in fraction and non-fracture porous media. Among their considerations were the choice of microbes limited to Bacillus substilis (a biosurfactant producing microbe) and leuconostomesenteroides (an exopolymer producing microbe). Their result showed that higher oil recovery efficiency will be achieved by using a biosurfactant producing bacteria in fractured porous media. The presentation of five successful MEOR projects was reported in [9]. These projects reflected diversity of locations, depth, porosity, permeability and temperatures which were conducted in different parts of the world. The projects included sandstone, fractured dolomite and fractured sandstone reservoirs. Reservoir depth ranged from 4450 to 6900ft, temperatures from 110°F to 180°F, porosity range from 0.079 to 0.232 and effective permeability from 17 to 300mD. Analysis of the above showed that MEOR recovers oil and reduced water production from highly permeable zones.

The discovery of soil samples to permeate fluid in different rates depending on the characteristics of the medium through which fluid flows was presented in [10]. He presented mathematically the rate at which fluid flows directly proportional to the area of the media through which the fluid is flowing. He outlined that for a MEOR performance, some properties are of great importance which includes; reservoir geometry, depth, pay uniformity and continuity, formation transmissibility, formation storage capacity and oil viscosity. The performance of a MEOR project and penetration of microbes through the formation greatly depends on formation orientation and rock grain arrangement in the reservoir. Plugging and clogging of the formation are some of the frequently encountered risks in MEOR...
application, the analysis of these effects on fluid flow, microbial concentration profile and production rates of recoverable oil forms the basis of this study.

II. INTRODUCTION

Conventional oil recovery methods cannot recover all of the oil in the reservoir when the reservoir has fully depleted its primary energy drive. Hence the need for microbial enhanced oil recovery (MEOR) which is a sophisticated technique meant for recovering high viscosity crude in the reservoir. Microbial enhanced oil recovery usually termed oil bio-refining and biodegradation results in recovering heavier oil due to its viscosity reduction effects, volume displacement, surface tension reduction, acid reaction, increased permeability, improved sweep efficiency and increased reservoir pressure [1], [2]. The matrix of a reservoir formation consists of organic and inorganic materials that are under compaction in a reservoir [3], [4]. It is also the arrangement of grains in the formation, consisting of organic and inorganic materials that are under compaction in a reservoir [4], [5]. The arrangement of the rock grains determines the permeability, porosity, inter-connectivity, tortuosity and fluid saturation of the formation.

III. RESEARCH METHODOLOGY

The mathematical model to be proposed for demonstrating bacteria transportation through porous media is based on a combination of the theories proposed by Gruesbeck et al [11] for description of entrainment and deposition of fines in porous media.

\[
\frac{\partial}{\partial t} \left[ \frac{\phi_{Sw}}{B_w} C_b + \phi_{Sw} b \right] = \nabla \left[ \frac{U}{B_w} C_b \right] + \nabla \left[ \frac{\phi_{Sw}}{B_w} D_{bw} \nabla C_b \right] - \frac{Q_w C_b}{V_p} + \frac{\phi_{Sw}}{B_w} R_b
\]

Removing Gradient operators and re-writing equation (1) in 3-D gives:

\[
\frac{\partial}{\partial t} \left[ \frac{\phi_{Sw}}{B_w} C_b + \phi_{Sw} b \right] = -\frac{\partial}{\partial x} \left( \frac{U}{B_w} C_b \right) + \left( \frac{\phi_{Sw}}{B_w} D_{bw} \frac{\partial C_b}{\partial x} \right) - \frac{Q_w C_b}{V_p} + \frac{\phi_{Sw}}{B_w} R_b
\]

\[
-\frac{\partial}{\partial y} \left( \frac{U}{B_w} C_b \right) + \left( \frac{\phi_{Sw}}{B_w} D_{bw} \frac{\partial C_b}{\partial y} \right) - \frac{Q_w C_b}{V_p} + \frac{\phi_{Sw}}{B_w} R_b
\]

\[
-\frac{\partial}{\partial z} \left( \frac{U}{B_w} C_b \right) + \left( \frac{\phi_{Sw}}{B_w} D_{bw} \frac{\partial C_b}{\partial z} \right) - \frac{Q_w C_b}{V_p} + \frac{\phi_{Sw}}{B_w} R_b
\]
Assuming single-phase fluid flow in 1-dimension (i.e X-direction) gives:

$$\frac{\partial}{\partial t} \left[ \frac{\phi S_w}{B_w} C_b + \phi C_{bw} \right] = -\frac{\partial}{\partial x} \left( \frac{U_t}{B_w} C_b \right) + \left[ \phi \frac{S_w}{B_w} D_{bw} \frac{\partial C_b}{\partial x} - \frac{Q_w C_b}{V_p} + \phi \frac{S_w}{B_w} R_b \right]$$

(3)

Where:
- $C_b$ = flowing concentration of the bacteria and substrate (lb/cft)
- $C_{bw}$ = Adsorbed concentration of bacteria and substrate at the pore surfaces (lb/cft)
- $S_w$ = Saturation of the water phase (%)
- $B_w$ = Formation volume factor of the water phase (cft/scft)
- $Q_w$ = Volumetric injection rate of bacteria and substrate via the water phase (stb/day)
- $\phi$ = Rock porosity (%)
- $V_p$ = pore volume of the porous medium
- $U_t$ = The total flow velocity (ft/day)
- $D_{bw}$ = physical dispersion of bacteria in the water phase (ft$^2$/day)
- $R_b$ = Biological bacteria growth (day$^{-1}$)

Where

$$\left[ \phi \frac{S_w}{B_w} c_e \right]$$

Represents the accumulation of microbes and substrates in the aqueous phase.

$(\phi C_{bw})$ denotes the adsorption of the bacteria and substrate in the pore space. It is a function of the rate of detachment and retention respectively.

The term $$\frac{\partial}{\partial x} \left( \frac{U_t}{B_w} C_b \right)$$ represents the convection term. The velocity ($U_t$) in this term is defined as:

$$U_t = U_w + K_c \frac{\partial}{\partial x} \left[ \frac{1}{C_s} \frac{\partial C_s}{\partial x} \right]$$

(4)

Where
- $U_w$ = Darcy Velocity (flux) for the water phase (ft/day)
- $K_c$ = chemotactic co-efficient (ft$^2$/day)
- $C_s$ = Concentration of substrate (lb/cft)

Chemotactic movement is defined as the directed movement of a cell towards a substrate. Microbes can sense a nutrient-rich environment and move in that direction.

While Darcy flow occurs due to pressure gradient, chemotactic migration of bacteria is assumed to be proportional to an exponential change in nutrient concentration. It is worthy of note, that chemotactic flow of bacteria is much smaller than convective flow and hence, it is significant only near static conditions. Therefore equation (4) becomes

$$U_t = U_w$$

(5)

1. The term $$\frac{\partial}{\partial x} \left[ \phi \frac{S_w}{B_w} V_{bw} \frac{\partial C_b}{\partial x} \right]$$ is the dispersion tensor. The elements of the dispersion tensor include both molecular diffusion and mechanical dispersion. Since the mechanical dispersion is mostly negligible, then the elements are given as:

$$D_{bw} = \frac{D_b}{\tau}$$

(6)

Where
- $D_b$ = Molecular diffusion co-efficient for bacteria and substrate in the water phase ( ft$^2$/ day )
- $\tau$ = Tortuosity of the porous media/reservoir (dimensionless)

2. The term $$\frac{Q_w C_b}{V_p}$$ is the microbial and substrate injection rate

3. The term $$\phi \frac{S_w}{B_w} R_b$$ is the bacteria reaction rate.
Bacterial growth can either occur in a single – substrate or a double substrate medium; though growth can be inhibited by alcohol or a metabolic product. The dependence of bacterial growth rate, $\mu_m$, on substrate concentration is represented by Monod’s equation

$$
\mu_m = \frac{\mu_{max} C_S}{K_s + C_S}
$$

(7)

Where

$\mu_{max}$ = Maximum specific growth rate obtained in excess substrates (day$^{-1}$)

$C_s$ = the concentration of growth limiting substrate (lb/cft)

$K_s$ = The substrate concentration corresponding to half $\mu_{max}$ (lb/cft). In general, $K_s$, for most growth substrates is very small.

Since there was no adsorption of bacteria, substitution of equations (5), and (7) into equation (3) yields

$$
\frac{\partial}{\partial t} \left( \phi \frac{S_w}{B_w} C_b \right) = - \frac{\partial}{\partial x} \left( U_w \frac{\partial C_b}{B_w} \right) + \frac{\partial}{\partial x} \left( \phi \frac{S_w}{B_w} D_{bw} \frac{\partial C_b}{\partial x} \right) - \frac{Q_w C_b}{V_p} + \phi \frac{S_w}{B_w} \mu_m
$$

(8)

If we factor out the term $\phi \frac{S_w}{B_w}$ and divide each term by it; we have

$$
\frac{\partial C_b}{\partial t} = - \frac{1}{\phi} \frac{U_w}{B_w} \frac{\partial C_b}{\partial x} + \frac{\partial}{\partial x} \left( \phi \frac{S_w}{B_w} D_{bw} \frac{\partial C_b}{\partial x} \right) - \frac{Q_w C_b}{V_p} \frac{1}{\phi} \frac{S_w}{B_w} + \mu_m
$$

(9)

Let $D_s = \phi \frac{S_w}{B_w}$ = formation water factor

Then equation 9 becomes

$$
\frac{\partial C_b}{\partial t} = - \frac{1}{D_s} \frac{U_w}{B_w} \frac{\partial C_b}{\partial x} + \frac{\partial}{\partial x} \left( \frac{D_{bw}}{D_s} \frac{\partial C_b}{\partial x} \right) - \frac{Q_w C_b}{V_p} \frac{1}{D_s} + \mu_m
$$

(10)

Equation (10) above is the model equation for describing microbial and substrate transport in porous media.

a. **Finite Difference Approximation**

FDA proves the best solution to the proposed microbial model, its description presented below;

Fig. 2 describes the spatial presentation of the finite difference approximation ranging from i-1 to i+1 respectively.
Fig. 3: Discretization system for a MEOR subjected Porous media

Fig. 3 illustrates a block centered grid system which was selected for the finite difference formulation. The X and Y directions are the areal coordinates and the positive z-direction is normal to the bedding plane in the downward direction. The blocks are numbered in natural order.

The model equation for microbial and substrate transport in porous media in equation (10) above can be written as:

$$\left( \frac{\partial C_b}{\partial t} \right)_i = -\left( \frac{U_w}{D_B B_w} \right)_i \left( \frac{\partial C_b}{\partial x} \right)_i + (D_{bw})_i \left[ \frac{\partial^2 C_b}{\partial x^2} \right]_i - \left[ \frac{\partial^2 C_b}{D_B V_p} \right]_i + [\mu_m]$$

(11)

Re-writing equation (11) using forward difference gives

$$\left( \frac{\partial C_b}{\partial t} \right)_i = \frac{C_{bi}^{n+1} - C_{bi}^n}{\Delta t}$$

(12)

$$\left( \frac{\partial C_b}{\partial x} \right)_i = \frac{C_{bi}^{n+1} - C_{bi}^n}{(\Delta x)}$$

(13)

$$\left( \frac{\partial^2 C_b}{\partial x^2} \right)_i = \frac{C_{bi}^{n+1} - 2C_{bi}^n + C_{bi-1}^n}{(\Delta x)^2}$$

(14)

Substituting equation (12), (13) and (14) into (11) gives

$$\left( \frac{\partial C_b}{\partial t} \right)_i = \left( \frac{U_w}{D_B B_w} \right)_i \left[ C_{bi}^{n+1} - C_{bi}^n \right]_i + (D_{bw})_i \left[ \frac{C_{bi}^{n+1} - 2C_{bi}^n + C_{bi-1}^n}{(\Delta x)^2} \right]_i - \left( \frac{Q_w C_{bi}}{D_B V_p} \right)_i + [\mu_m]$$

(15)

b. Spatial Discretization

At a grid-block boundary, the first term on the RHS of equation (10) can be re-written using finite difference approximation for spatial derivatives as follows:

$$\left[ \frac{U_w}{D_B B_w} \right]_i \left[ \frac{\partial C_b}{\partial x} \right]_i = \left[ \frac{U_w}{D_B B_w} \right]_{i+\frac{1}{2}} \left[ \frac{\partial C_b}{\partial x} \right]_{i+\frac{1}{2}} - \left[ \frac{U_w}{D_B B_w} \right]_{i-\frac{1}{2}} \left[ \frac{\partial C_b}{\partial x} \right]_{i-\frac{1}{2}}$$

(16)
\[
\left[ \frac{\partial C_{bi}}{\partial x} \right]_{i+\frac{1}{2}} = \frac{C_{bi+1} - C_{bi}}{\Delta X_{i+\frac{1}{2}}}
\]

And
\[
\left[ \frac{\partial C_{bi}}{\partial x} \right]_{i-\frac{1}{2}} = \frac{C_{bi} - C_{bi-1}}{\Delta X_{i-\frac{1}{2}}}
\]

Substitution of equation (17) and (18) into (16) gives
\[
\left[ \frac{U_w}{D_s B_w} \right] \left[ \frac{\partial C_{bi}}{\partial x} \right]_{i+\frac{1}{2}} = \left[ \frac{U_w}{D_s B_w} \right] \left( \frac{C_{bi+1} - C_{bi}}{\Delta X_{i+\frac{1}{2}}} \right) - \left( \frac{C_{bi} - C_{bi-1}}{\Delta X_{i-\frac{1}{2}}} \right)
\]

Assuming uniform grid-block size and spacing we have
\[
\left[ \frac{U_w}{D_s B_w} \right] \left[ \frac{\partial C_{bi}}{\partial x} \right]_{i+\frac{1}{2}} = \left[ \frac{U_w}{D_s B_w} \right] \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right)
\]

Substitution of equation (20) into equation (15) yields
\[
\frac{C_{bi}^{n+1} - C_{bi}^n}{\Delta t} = -\left[ \frac{U_w}{D_s B_w} \Delta X_{i+\frac{1}{2}} \right] \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right) \left[ \frac{D_{bw}}{D_s B_w} \left( \frac{\Delta \mu}{\mu} \right) \right] - \frac{Q_w C_{bi}^n}{D_s V_p} + \mu_m
\]

If we assume \( \Delta X_{i} = \Delta X_{i+\frac{1}{2}} \) (i.e. uniform block boundary and grid block size / spacing), then
\[
\frac{C_{bi}^{n+1} - C_{bi}^n}{\Delta t} = \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right) \left[ \frac{D_{bw}}{D_s B_w} \left( \frac{\Delta \mu}{\mu} \right) \right] - \frac{Q_w C_{bi}^n}{D_s V_p} + \mu_m
\]

c. Explicit Formulation

The explicit formulation of the model in (22) assumes a base time level of n. This implies that the microbial concentrations on the RHS of (22) assumes the n-base time level \( t^n \) and therefore can be re-written as
\[
C_{bi}^{n+1} - C_{bi}^n = \Delta t \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right) \left[ \frac{D_{bw}}{D_s B_w} \left( \frac{\Delta \mu}{\mu} \right) \right] - \frac{Q_w C_{bi}^n}{D_s V_p} + \Delta t \mu_m
\]

Resolving (23), we obtain
\[
C_{bi}^{n+1} - C_{bi}^n = \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right) \left[ \frac{\Delta t D_{bw}}{D_s B_w \Delta X} \right] - \frac{Q_w \Delta t}{D_s V_p} C_{bi} + \Delta t \mu_m
\]

The above yields
\[
C_{bi}^{n+1} = C_{bi}^n + \left( C_{bi+1} - 2C_{bi} + C_{bi-1} \right) \left[ \frac{\Delta t D_{bw}}{D_s B_w \Delta X} \right] - \frac{Q_w \Delta t}{D_s V_p} C_{bi} + \Delta t \mu_m
\]
Rearranging the above
\[ C_{bi}^{n+1} = C_{bi}^n - 2C_{bi}^n \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) Q_w \frac{\Delta t}{D_u V_p} C_{bi}^n + \left( C_{bi}^n - C_{bi}^{n-1} \right) \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) + \Delta \mu_m \] (26)

In terms of \( C_{bi}^{n+1} \), the above can be written as;
\[ C_{bi}^{n+1} = C_{bi}^n \left[ 1 - 2 \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) - \frac{Q_w \Delta t}{D_u V_p} \right] + \left( C_{bi}^n - C_{bi}^{n-1} \right) \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) + \Delta \mu_m \] (27)

\[ \dot{M} = \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) \] (28)

Let
\[ \dot{M} = \left( \frac{\Delta D_{bw}}{(\Delta x)^2} - \frac{\Delta U_w}{D_u B_w \Delta X} \right) \] (29)

The following assumptions were made in deriving the deduced model:
1) The bacteria and substrate were transported into the reservoir via the water phase.
2) There was no external production during the period of microbial injection.
3) There was no adsorption of bacteria on the pore spaces during propagation and therefore bacterial rates for retention and detachment were to considered
4) Minimal substrate metabolites production
5) Change sin porosity (\( \phi \)) phase saturations (\( S_w \)) and formation volume factors (\( B_w \)) were small compared to changes in concentrations.
6) Chemo-taxis. Not considered.
7) Effects of mechanical tensor were negligible.
8) Gravitational effects were neglected.
9) Microbial injection and transportation occurred in a homogenous and isotropic reservoir (porous media) i.e. \( K \) = constant at constant velocity (\( U_w \))
10) Area of microbial flow in porous media is constant.
11) There was no microbial influx at boundaries during propagation.
12) Other factors affecting MEOR such as salinity and pH were not considered.
13) No specific microbial consideration

IV. RESULTS AND DISCUSSION

Using field parameters from X field, the rock and fluid properties are presented below
Formation porosity \( \phi = 20\% \);
Reservoir Dimension \( \Delta x = 1000ft; \Delta y = 500ft; \Delta z = 200ft \)
\( k_h = K_v = K_z = 100md, \)
\( P = 2500 \text{ Psia} \)
\( S_w = 0.3 \)
\( B_w = 1 \text{ rb/stb} \)
\[ \mu_w = 1.0 \text{cp} \]
\[ Q_w = 100 \text{ bbl/day} \]

Also,

- Bulk volume \( (V_b) = \Delta x \cdot \Delta y \cdot \Delta z = (100 \times 500 \times 200) \text{ ft}^3 = 100 \times 10^6 \text{ ft}^3 \)
- Area of reservoir \( = \Delta x \cdot \Delta y = (1000 \times 500 \times 200) \text{ ft}^3 = 100,000 \text{ ft}^2 \)

Velocity of injected water,
\[ U_w = \frac{K \Delta P}{\mu_w \Delta x} = \frac{100 \times 2500}{1 \times 1000} = 250 \times 1.127 \times 10^{-3} = 0.28175 \times 5.615 = 1.582 \text{ ft/day} \]

Assumed Parameters:
- Initial microbial concentration, \( C_{bi} = 300 \text{ lb/ft}^3 \)
- Bacteria growth rate (Monod’s constant), \( \mu_p = 1.3 \text{ Day}^{-1} \)
- Molecular Diffusion of bacteria, \( D_b = 0.2 \text{ ft}^2/\text{Day} \)
- Tortuosity of formation (Sandstone) \( = 2 \)
- Time-step (interval of iteration) \( = 5 \text{ days} \)

Calculating constants
\[ D_{bw} = \frac{D_w}{t} = \frac{0.2}{2} = 0.1 \text{ ft}^2/\text{Day} \]
\[ D_s = \frac{\phi S_w}{B_w} = \frac{0.2 \times 0.3}{1} = 0.06 \]
\[ \dot{M} = \frac{\Delta t \cdot D_{bw}}{(\Delta x)^2} = \frac{U_w \Delta t \Delta x}{(1000)^2} \left( \frac{5 \times 0.1}{(1.582 \times 5)} \right) \left( \frac{0.06 \times 1 \times 1000}{1} \right) = 1.0 \times 10^{-6} - 0.2637 = -0.1318 \]

Table 1 above displays average microbial concentration at various time intervals across the reservoirs viz: homogenous and heterogeneous (using averaged and non-averaged permeabilities).
Fig. 4 shows a plot of average microbial concentration against time in Fig. 4 gives a straight line. The implication of this is that for a homogenous reservoir system having constant permeability and porosity, there is a constant increase in biomass concentration at every point in time. Since the formation matrixes are uniformly arranged, transportation of microbes were not breached, distorted or impeded. This results in a steady and defined transportation of the injected microbes.

Fig. 5: plot of microbial concentration against time for heterogenous averaged permeability

Fig. 5 presents the response of the concentration of microbes with increasing time for a heterogeneous reservoir system, adopting the averaging method for porosity and permeability. It is observed that for varying formation matrix parameters such as porosity and permeability, the microbial concentration profile is distorted. Therefore the average microbial concentration trend with increasing time will not be uniform as seen for the homogeneous system.
Biomass concentration per unit time was also calculated for heterogeneous reservoir whose varying rock parameters without adopting the averaging method for porosity and permeability. The inconsistencies observed in the plots of concentration against time with different porosity and permeability is shown in Fig 6. A reduction in the values of microbial concentration was observed as microbes moved across the reservoir grid blocks from injector well to the producer, traceable to the clogging tendency of the microbes as they meander through the formation. The above phenomenon will result in an inefficient microbial oil recovery process.

A comparison of concentration of the injected microbes with respect to injection time is presented for homogeneous, heterogeneous and heterogeneous Averaged permeability in Fig 7. Both heterogeneous formations appear to follow the same trend, but the non-averaged permeability formation recorded a lower microbial concentration over the injection period.

V. CONCLUSION

MEOR is almost applicable to any type of reservoir with respect to their depositional characteristics, reservoir configuration, bed arrangement etc. Adaptation of the averaging method for both permeability and porosity will
minimize the degree of distortion in the concentration profile of the injected microbe. The performance efficiency of MEOR will be higher in magnitude for a homogenous reservoir as when compared to the performance of MEOR for heterogeneous reservoirs, irrespective of the adaptation of the averaging method for the rock properties. It is understood that variation in rock properties such as porosity and permeability will significantly determine the transport, mobility, propagation, performance and efficiency of the MEOR process.

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