

# Quasi-Newton Methods for Solving Nonlinear Programming Problems

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## Abstract

In the present paper the problem of constrained equality optimization is reduced to sequential solving a series of problems of quadratic programming. The Hessian of the Lagrangian is approximated by a sequence of symmetric positive definite matrices. The matrix approximation is updated at every iteration by a Gram-Schmidt modified algorithm. We establish that methods is locally convergent and the sequence  $\{x_k\}$  converges to the solution a two-step superlinear rate.

Key words: Quasi-Newton methods, Constrained Optimization, Superlinear convergence.

## 1 Introduction

This paper considers the methods of the finding of a vector  $x_* \in R^n$ , which is the solution of the nonlinear programming problem:

$$\begin{aligned} & \text{minimize } f(x) \\ & \text{subject to } h_i(x) = 0, \quad 1 \leq i \leq r. \end{aligned} \quad (1)$$

It is supposed that the following conditions are fulfilled:

- 1) the functions  $f(x) : R^n \rightarrow R^1$ ,  $h_i(x) : R^n \rightarrow R^1$  are continuously differentiated twice in some neighbourhood of the point  $x_*$ ;
- 2) the gradients  $\nabla h_1(x_*), \nabla h_2(x_*), \dots, \nabla h_r(x_*)$  form a linear independent sistem;

- 3)  $\nabla_x L(x_*, \lambda_*) = 0$ , where  $\nabla_x L(x, \lambda)$  is a vector of the first derivatives by  $x$  from the Lagrangian function

$$L(x, \lambda) = f(x) + \sum_{i=1}^r \lambda^i h_i(x);$$

- 4) the matrix  $\nabla_{xx}^2 L(x_*, \lambda_*)$  is positive definite on the null space of the constraint gradients, i.e.

$$\eta^T \nabla_{xx}^2 L(x_*, \lambda_*) \eta \geq m \|\eta\|^2, 0 < m < \infty,$$

at any vector  $\eta \in R^n, \eta \neq 0$ , such as  $\nabla h(x_*) \eta = 0$ , where  $\nabla_{xx}^2 L(x, \lambda)$  is a matrix of the second derivatives of the Lagrangian function  $L(x, \lambda)$  by  $x$ ;  $\nabla h(x)$  is a matrix, the lines of which are the gradients  $\nabla h_i(x), 1 \leq i \leq r$ . The symbol "T" denotes the transpose of a vector or a matrix.

If the conditions 1) to 4) are fulfilled  $x_*$  is the point of the strict local minimum of the function  $f(x)$  with the restrictions (1) (see for example [1], p.42-49). Besides that, let us apply the Newton's method to the system of the nonlinear equations (see [2], p.275):

$$\left. \begin{array}{l} \nabla_x L(z) = 0, \\ h_1(x) = 0, \\ \dots\dots\dots \\ h_r(x) = 0. \end{array} \right\} \quad (2)$$

Here  $z = (x, \lambda)^T$ . Throughout the paper we will denote by  $z_k = (x_k, \lambda_k)^T$  and by  $z_* = (x_*, \lambda_*)^T$  the Kuhn-Tucker pair for problem (1).

The process, caused by the Newton's method, while applying it to the system (2) is equivalent (see [3-8]) to the solution of the quadratic programming problem of minimization a the function

$$Q(x, \lambda_k) = \frac{1}{2} (x - x_k)^T \nabla_{xx}^2 L(z_k) (x - x_k) + \nabla f(x_k)^T (x - x_k) \quad (3)$$

subject to the linear constraints

$$h_i(x_k) + \nabla h_i(x_k)^T(x - x_k) = 0, \quad 1 \leq i \leq r. \quad (4)$$

Here  $\lambda_{k+1}$  will be the vector of the Lagrange's multipliers in the extreme point  $x_{k+1}$  of the problem (3),(4), i. e. that is  $x_{k+1}$  and  $\lambda_{k+1}$  are calculated by  $x_k$  and  $\lambda_k$  as a solution and a vector of the Lagrange's multipliers of the problem (3),(4) connected with it. If the initial approximation  $z_0$  is chosen from the sufficiently small neighbourhood of the point  $z_*$ , the sequence  $x_0, x_1, \dots, x_k, \dots$  converges to  $x_*$  and besides that it converges with a superlinear rate (see more precise result in [6,9,10]).

One of the main disadvantages of this method is that the function's  $Q(x, \lambda_k)$  convexity is not guaranteed. Another one is that while determining every new approximation it is necessary to calculate the Hesse matrix of the Lagrangian function  $L(x, \lambda)$  at the point which corresponds to the precedent approximation. That is why a question appears how to construct the methods, which do not require the calculation of the matrix  $\nabla_{xx}^2 L(x_k, \lambda_k)$  and provide the convexity of the auxiliary problems (3),(4) keeping the high velocity of convergency. At present, such a method is elaborated, based on using of the variable metric algorithms (see [6,11–15]). In this papers the matrices  $\nabla_{xx}^2 L(x_k, \lambda_k), k = 0, 1, 2, \dots$ , are replaced by some other matrices  $A_k$ , which are positively definite on the space  $R^n$  and approximate the matrix  $\nabla_{xx}^2 L(x_*, \lambda_*)$  only on the tangent subspace to the constraints of the problem (1) in the point  $x_*$ . The construction of the matrices  $A_k, k = 0, 1, 2, \dots$ , are accomplished with help of the variable metric algorithms.

In our work we present methods, in which on each iterates the convex problems of the quadratic programming are solved and the approximation of the required matrices is accomplished by the methods, examined in the papers [16–18] will be studied.

## 2 The method and its proprieties

Consider a matrix

$$M(z) = \nabla_{xx}^2 L(z) + \nabla h(x)^T D \nabla h(x), \quad (5)$$

where  $D$  is some bounded symmetric and positively definite matrix of the dimension  $r \times r$ .

The matrix  $\nabla h(x)^T D \nabla h(x)$  is positively definite on the set

$$\{y \in R^n : \nabla h(x)y \neq 0\}.$$

According to the assumptions done in Section 1 it is possible to select a matrix  $D$  such, that the matrix  $M(z)$ , which is defined by formula (5) would be positively definite on  $R^n$  in some neighbourhood of the point  $z_*$ . More detailed about the selection of the matrix  $D$  will be said later in Section 3.

Further, every where will consider, that the matrix  $D$  is selected such that

$$y^T M(z)y \geq m \|y\|^2, \forall y \in R^n, m > 0, \quad (6)$$

for all  $z$  from the small enough neighbourhood of the point  $z_*$ , marked by  $\Omega$ .

Let  $z_0 \in \Omega$ . Consider an iteration process in which  $z_{k+1}$  is defined according to  $z_k$  from the solution of the problem of minimization of the quadratic function

$$\tilde{Q}(x, \lambda_k) = \frac{1}{2}(x - x_k)^T M(z_k)(x - x_k) + \nabla f(x_k)^T (x - x_k) \quad (7)$$

with the constraints (4).

By the solution of the quadratic programming problem (7), (4) it is understood the same as for the problem (3), (4). This process of the construction of the sequence  $\{z_k\}$  will name further as the method (7), (4).

**Lemma 1** *The sequence  $\{z_k\}$  constructed according to the method (7), (4) converges to  $z_*$  at superlinear rate.*

*Proof* is done analogically to the proof of the Newton's method of solving the systems of equation [19]. In the considered case

$$z_{k+1} = z_k - [B(z_k)]^{-1}F(z_k), \quad (8)$$

where

$$B(z_k) = \begin{bmatrix} M(z_k) & \nabla h(x_k)^T \\ \nabla h(x_k) & 0 \end{bmatrix}$$

and

$$F(z_k) = \begin{bmatrix} \nabla_x L(z_k) \\ h_1(x_k) \\ \dots\dots\dots \\ h_r(x_k) \end{bmatrix}$$

The matrix  $B(z_k)$  is nonsingular according to the assumptions 1)–4) from Section 1. It is possible to prove that in the same way as it was done in [2], p.275. Hence, on  $\Omega$  is correctly defined the operator

$$G(z) = z - [B(z)]^{-1}F(z) \quad (9)$$

and besides

$$G'(z_*) = I - [B(z_*)]^{-1}\mathfrak{S}(z_*),$$

where  $I$  is unit matrix of the dimension  $(n+r) \times (n+r)$  and  $\mathfrak{S}$  is the  $(n+r) \times (n+r)$  Jacobian matrix of  $F$ .

It is easy to convinced (see [20] p. 252)

$$[B(z)]^{-1} = [\mathfrak{S}(z)]^{-1} - \begin{bmatrix} 0 & 0 \\ 0 & D \end{bmatrix} \quad (10)$$

Taking this in consideration we'll obtain:

$$G'(z_*) = \begin{bmatrix} 0 & 0 \\ D\nabla h(x_*) & 0 \end{bmatrix}$$

that is the matrix  $G'(z_*)$  is a triangular matrix, and it means that it's eigenvalues  $\mu_1, \mu_2, \dots, \mu_{n+r}$  exactly coincide with the diagonal elements. But  $r \leq n$ , because in the opposite case the system of vector  $\{\nabla h_i(x_*)\}_{i=1}^r$  will not be linearly independent. Hence it follows, that

$$\mu_1 = \mu_2 = \dots = \mu_{n+r} = 0.$$

According to the definition of the spectral radius

$$\rho(G'(z_*)) = \max\{|\mu_1|, \dots, |\mu_{n+r}|\} = 0. \quad (11)$$

Now to finish the proof of the lemma, it is necessary to use the theorem of Ostrowsky, which proves: if  $\rho(G'(z_*)) < 1$  then the  $G(z)$  which is defined by formula (9) is a contraction. The process (8) is nothing else than a method of successive approximation for the solution of the equations  $G(z) = z$ , that is finding of immovable point of the operator  $G(z)$ .

The equality (11) guarantees that the sequence  $\{z_k\}$  converges to the point  $z_*$ , more rapidly than any geometrical progression. Indeed, for  $\forall$  given  $\varepsilon/2 > 0$  exist a norma  $\|\bullet\|$  in  $E^n$  (see, for example, [19], p. 47) such that

$$\|G'(z_*)\| \leq \rho(G'(z_*)) + \frac{\varepsilon}{2} = \frac{\varepsilon}{2},$$

according to (11). For arbitrary  $\varepsilon/2 > 0$  according to differentiation of  $G(z)$  it is possible to indicate a number  $N = N(\varepsilon)$  such that

$$\|G(z_k) - G(z_*) - G'(z_*)(z_k - z_*)\| \leq \frac{\varepsilon}{2} \|z_k - z_*\|,$$

when  $k \geq N$ . Taking this in consideration, for  $k \geq N$

$$\begin{aligned} \|z_{k+1} - z_*\| &\leq \|G(z_k) - G(z_*) - G'(z_*)(z_k - z_*)\| + \\ &+ \|G'(z_*)\| \|z_k - z_*\| \leq \varepsilon \|z_k - z_*\|. \end{aligned}$$

Lemma is proved.

*Remark.* If  $h_i(x), 1 \leq i \leq r$  are the linear functions then  $x_k \rightarrow x_*$  superlinearly. In fact, in this case from (8) taking in consideration (10) will have

$$z_{k+1} = z_k - [F'(z_k)]^{-1}F(z_k),$$

that is the matrix  $\nabla h(x_k)^T D_k \nabla h(x_k)$  does not participate in the construction of the vector  $z_{k+1}$ . In the common case it is impossible to get a superlinear convergence for the sequence  $\{x_k\}$ .

The construction of the vector  $x_{k+1}$ , which is the solution of the problem (7), (4) does not depend on the selection of the matrix  $D_k$ . This matrix influences sufficiently only on the properties of the sequence  $\{\lambda_k\}$ . It is possible to show that in the following way. Let the point  $z_k$  be constructed and let  $\bar{x}_{k+1}, \bar{\lambda}_{k+1}$  be the solution of the problem (3), (4). Taking in consideration (10) it's easy to see, that the solution of the problem (7), (4) is connected with  $\bar{x}_{k+1}, \bar{\lambda}_{k+1}$  in the following way:  $x_{k+1} = \bar{x}_{k+1}, \lambda_{k+1} = \bar{\lambda}_{k+1} - D_k h(x_k)$ . From this follows, that the selection of the matrix  $D_k$  directly shows up when constructing vector  $\lambda_{k+1}$ . This selection finally will influence on the properties of the sequence  $\{x_k\}$  because the value of  $\lambda_k$  is required while forming the problems (7), (4). The character of convergence is set in the following lemma.

**Lemma 2** *For the sequence  $\{x_k\}$ , constructed by method (7), (4) the following appraisal is true*

$$\|x_{k+1} - x_*\| \leq \mu_k \|x_{k-1} - x_*\|, \quad (12)$$

where  $\mu_k \rightarrow 0$ , while  $k \rightarrow \infty$ .

*Proof.* Let

$$P_k = I - \nabla h(x_k)^T [\nabla h(x_k) \nabla h(x_k)^T]^{-1} \nabla h(x_k).$$

In the paper [6] the following statement is set if the matrix  $M(z_k)$  in the method (7), (4) is such that

$$\lim_{k \rightarrow \infty} \frac{\|P_k(M(z_k) - \nabla_{xx}^2 L(z_*))P_k(x_{k+1} - x_k)\|}{\|x_{k+1} - x_k\|} = 0 \quad (13)$$

then for the sequence  $\{x_k\}$  the appraisal (12) is true. To prove the lemma it is enough to show, that for the matrix  $M(z_k)$ , selected in the form (5) the condition (13) is true.

In fact for any vector  $y \in E^n$

$$\nabla h(x_k)P_k y = 0.$$

Hence,

$$(M(z_k) - \nabla_{xx}^2 L(z_k))P_k y = 0.$$

Taking this in consideration, after some transfofmations, we obtain

$$\begin{aligned} & \|P_k(M(z_k) - \nabla_{xx}^2 L(z_*)P_k(x_{k+1} - x_k))\| = \\ & = \|P_k(\nabla_{xx}^2 L(z_k) - \nabla_{xx}^2 L(z_*))P_k(x_{k+1} - x_k)\| \leq \gamma_k \|x_{k+1} - x_k\|, \end{aligned}$$

where

$$\gamma_k = \|\nabla_{xx}^2 L(z_k) - \nabla_{xx}^2 L(z_*)\| \rightarrow 0,$$

while  $k \rightarrow \infty$ , because  $z_k \rightarrow z_*$ , according to lemma I.

### 3 The Description of the Quasi-Newton Algorithms

The properties of the method (7), (4) will be kept, if the matrix  $M(z_k)$  will be replaced by some other matrix  $A_k$ , which is sufficiently close to the matrix  $M(z_k)$ . Let's pass now to the construction of the matrices  $A_k$  (which approximate  $M(z_k)$ , using the methods studied in [18].

It's necessary to note that the problem of construction of these matrices is complicated, because obviously there does not exist function, the matrix of the second derivatives of which (calculated in the point  $z_k$ ) would be exactly equal to  $M(z_k)$  because the second addendum in



the matrix  $M(z_k)$  contains only the first derivatives of the functions  $h_i(x), 1 \leq i \leq r$ .

However one can take the function

$$\Psi(z) = L(z) + \frac{1}{2} \|\sqrt{D}h(x)\|^2.$$

In fact

$$\begin{aligned} \nabla_x \Psi(z) &= \nabla_x L(z) + \nabla h(x)^T D h(x), \\ \nabla_{xx}^2 \Psi(z) &= \nabla_{xx}^2 L(z) + \nabla h(x)^T D \nabla h(x) + h''(x)^T D h(x). \end{aligned}$$

Near the solution  $x_*$  it is possible to neglect the matrix  $h''(x)^T D h(x)$ . Hence, in the formulae (3.1), (4.1), (5.1), (6.1), (6.3) from the paper [18], for construction of the matrices  $A_k$ , it is possible to select as  $e_{k-i}$ , a vector

$$e_{k-i} = \nabla_x \Psi(x_{k-i} + r_{k-i}, \lambda_k) - \nabla_x \Psi(x_{k-i}, \lambda_k).$$

Consider another more simple method of construction of the vectors  $e_{k-i}, 1 \leq i \leq n$ , than the method above. Let  $e_{k-i}$  be calculated according to the formulae

$$\begin{aligned} e_{k-i} &= \nabla_x L(x_{k-i} + r_{k-i}, \lambda_k) - \nabla_x L(x_{k-i}, \lambda_k) + \\ &+ \nabla h(x_k)^T D_{k-i} [h(x_{k-i} + r_{k-i}) - h(x_{k-i})], 1 \leq i \leq n \end{aligned} \quad (14)$$

From the point of view of calculations it is more simple to take as  $D_{k-i}$  a product of the unit matrix  $I$  with a positive number  $\sigma_{k-i}$ :

$$D_{k-i} = \sigma_{k-i} I.$$

In this case

$$(e_{k-i}, r_{k-i}) = (a_{k-i}, r_{k-i}) + \sigma_{k-i} (b_{k-i}, \nabla h(x_k) r_{k-i}), \quad (15)$$

where

$$a_{k-i} = \nabla_x L(x_{k-i} + r_{k-i}, \lambda_k) - \nabla_x L(x_{k-i}, \lambda_k),$$

$$b_{k-i} = h(x_{k-i} + r_{k-i}) - h(x_{k-i}).$$

Two cases are possible on each  $i$ .

- 1)  $\nabla h(x_k)r_{k-i} = 0$ . According to the suppositions made above (see section 1) it means, that the vector  $r_{k-i}$  belongs to subspace, on which the matrix  $\nabla_{xx}^2 L(z_*)$  is positively defined.

Hence,  $(e_{k-i}, r_{k-i}) \geq m \|r_{k-i}\|^2$  on any selection of  $\sigma_{k-i}$ . Here,  $m$  — a constant that characterize (see section 1) a positive definite of the matrix  $\nabla_{xx}^2 L(z_*)$ .

- 2)  $\nabla h(x_k)r_{k-i} \neq 0$ . Using the Lagrange's formula for operators, it is possible to set, that the ratio  $(a_{k-i}, r_{k-i}) / (b_{k-i}, \nabla h(x_k)r_{k-i})$  is bounded. Let

$$\sigma_{k-i} = \delta - (a_{k-i}, r_{k-i}) / (b_{k-i}, \nabla h(x_k)r_{k-i}), \quad (16)$$

where  $\delta$  is an arbitrary positive number, chosen such that  $\sigma_{k-i} > 0$ .

Then from (15) we obtain

$$(e_{k-i}, r_{k-i}) = \delta (b_{k-i}, \nabla h(x_k)r_{k-i}) \quad (17)$$

While the gradients of the functions  $h_i(x), 1 \leq i \leq r$  are linearly independent in the neighbourhood of the points  $x_*$  the following appraisal is true

$$(e_{k-i}, r_{k-i}) \geq \delta m_1 \|r_{k-i}\|^2, 0 < m_1 < \infty.$$

Basing on that, it is possible now to construct the matrixes  $A_k$ , in the following way [18,21,22]. Suppose, that  $k = \xi n + s, \xi = 0, 1, 2, \dots, 0 \leq s \leq n - 1$ . We will denote  $w_{\xi n + s} = \eta_{\xi n + s} v_{s+1}$ , where  $v_1, v_2, \dots, v_n$  is a system of unit vectors. Let us calculate the matrix

$$A_k = \sum_{i=0}^{n-1} \frac{e_{k-i} e_{k-i}^T}{r_{k-i}^T e_{k-i}}. \quad (18)$$

Here the vectors  $e_{k-i}$  are chosen according to the formula (14), where

$$D_{k-i} = \sigma_{k-i} I,$$

where  $\sigma_{k-i}$  is either positive number or it is chosen from the condition (16).

The construction of the vectors  $r_{k-i}$  is performed according to the formulae

$$r_{\xi n} = w_{\xi n}, \quad r_k = w_k - \sum_{j=0}^{s-1} \frac{w_k^T e_{\xi n+j}}{e_{\xi n+j}^T r_{\xi n+j}} r_{\xi n+j}.$$

The factors  $\eta_k$  are chosen so that to satisfy the following condition  $0 < c \leq |\eta_{k+1}|/|\eta_k| \leq 1$  and  $\lim_{k \rightarrow \infty} \eta_k = 0$ . The vectors  $r_k, e_k$  are constructed recursively while using Gram-Schmidt modified algorithm [18,21,22].

Let's note, that if  $(a_{k-i}, r_{k-i}) \leq 0$  then it is possible to choose as  $\delta$  in (16) any positive number but the scalar product  $(e_{k-i}, r_{k-i})$  could be calculated directly by formula (17).

Analogically to that in [16,17] the verity of the following appraisals is set

$$(r_{k-i}, e_{k-j}) = o(\|r_{k-i}\| \|e_{k-j}\|), \quad i \neq j. \quad (19)$$

Taking in consideration these appraisals we'll have

$$A_k r_{k-i} = e_{k-i} + \mu_{k-i}, \quad 1 \leq i \leq n,$$

where  $\|\mu_{k-i}\| \rightarrow 0$ , while  $k \rightarrow \infty$ .

Repeating now the reasoning from the lemma 2.3.1. from [2], we'll set that

$$\lim_{k \rightarrow \infty} \|A_k - M(z_k)\| = 0 \quad (20)$$

If in addition the condition of Lipschitz for matrix  $\nabla_{xx}^2 L(x, \lambda)$  concerning  $x$  holds then the following appraisal is true:

$$\|A_k - M(z_k)\| \leq c|\eta_k|, 0 < c < \infty, \quad (21)$$

where  $\eta_k$  — a number used in the definition of the vectors  $r_{k-i}$ , (see [18]).

If the condition (19) holds true, then

$$\lim_{k \rightarrow \infty} \|(A_k - \nabla_{xx}^2 L(z_k))q\| = 0 \quad (22)$$

on any vector  $q$  such that  $\nabla h(x_k)q = 0$ .

Hence, the matrices  $A_k$  approximate  $\nabla_{xx}^2 L(z)$  on the tangent subspace to the multitude  $\{x | h_i(x) = 0, 1 \leq i \leq r\}$  at the point  $x_k$ , that is on that subspace, where  $\nabla_{xx}^2 L(z)$  is positive definite. Taking in consideration (20), (22), analogically to that as it was done in the proof of lemma 2, it is possible to show, that the considered Quasi-Newton methods will converge with the velocity, evaluated by inequality (12).

Let us mention the following. It is impossible to use the methods [18] for approximation of the matrix  $\nabla_{xx}^2 L(z_k)$  because it is possible that at some  $i$  will be  $(r_i, e_i) = 0$ . The exception is the method of indecisive directions [2]. While using this method, it is better to construct the matrices  $A_k$ , which approximate  $\nabla_{xx}^2 L(z_k)$  directly. To the constructed matrix  $A_k$  it is possible to add now a matrix  $\nabla h(x_k)^T D_k \nabla h(x_k)$ , trying to get a positive definite by choosing the matrix  $D_k$ .

It's necessary to mention that it is profitable to ensure the positive definite of required matrices, because in this case the auxiliary problems are rather effectively solved.

## 4 Another method of the constructing of the approximating matrices

Let's consider briefly one more method (that essentially differs from the considered above) of the construction of the matrices  $A_k$ , that ensures the subfilment of the condition (22).

According to the suppositions done above, the space

$$R(x_k) \equiv R_k = \{q | \nabla h(x_k)q = 0\}$$

has the dimensions  $n-r$ . Let  $b_1, \dots, b_{n-r}$  the basis of this space and let  $A_k$  is the matrix, defined by formula (18) in which  $w_{k-i} = \eta_{k-i}b_i, 1 \leq i \leq n-r$ , but the values  $e_{k-i}, 1 \leq i \leq n-r$  are replaced by the expressions  $a_{k-i}$  (see (15)) and addition is done by all  $i$  from 1 to  $n-r$ . Taking this constructions in consideration and reasoning in the same way as in [16–19] the verity of the appraisal (19) used, but then supposing that this appraisal holds true the verity of the condition (22) is proved. The matrices  $A_k$  satisfy the following condition  $(A_k y, y) \geq 0$  for any  $y \in E^n$ .

Concerning the construction of the basis of subspace  $R_k$  it could be obtained in the following way (see for example [23]). With the help of elementary operations and rearrangements the matrix  $\nabla h(x_k)$  is reduced to the matrix  $S_k$  strairs aspect. Then the system of equations  $\nabla h(x_k)q = 0$  is equivalent to the system  $S_k q = 0$ , which (taking in consideration that  $\text{rang}(\nabla h(x_k)) = r$ ) has  $n-r$  fred variables. Giving in turn to each of free variables the value 1, but to the rest of free variables the values 0 and solving the system  $S_k q = 0$  relative to the remained variables, we'll obtain search basis  $b_1, \dots, b_{n-r}$  of the space  $R_k$ .

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