



# 差分 化算法研究

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## Classic DE算法

初始化种群  $x_{i,1}, i=1,2,\dots,NP$  和变异因子  $F, CR$  以及  $j_{rand}$

$$x_{i,g} = \{x_{i,g}^1, x_{i,g}^2, \dots, x_{i,g}^D\}, i = 1, 2, \dots, NP, g = 1, 2, \dots, g_{max}$$

$$v_{i,g} = x_{r1,g} + F * (x_{r2,g} - x_{r3,g}), r1 \neq r2 \neq r3 \neq i$$

$$u_{i,g}^j = \begin{cases} v_{i,g}^j, & \text{如果 } (rand(0,1) \leq CR \text{ 或者 } j = j_{rand}) \\ x_{i,g}^j, & \text{否则} \end{cases}$$

$$u_{i,g}^j = L^j + rand(0,1) * (U^j - L^j)$$

$$x_{i,g+1} = \begin{cases} u_{i,g}, & \text{如果 } f(u_{i,g}) < f(x_{i,g}) \\ x_{i,g}, & \text{否则} \end{cases}$$



# JADE算法

$$[0,1] \quad \mu_F = (1 - c) \cdot \mu_F + c \cdot L_2(S_f) \quad \mu_F \in$$

~~$$\mu_{CR} = (1 - c) \cdot \mu_{CR} + c \cdot L_1(S_{cr}) \quad \mu_{CR} \in [0,1]$$~~

~~$$\mu = \frac{\sum_{k=1}^n z_k^p}{\sum_{k=1}^n z_k^{p-1}}$$~~



## jDE算法

$$F' = \begin{cases} 0.1 + rand1 * 0.9 & \text{如果 } rand0 < pi1 \\ F & \text{否则} \end{cases}$$

$$CR' = \begin{cases} rand3 & \text{如果 } rand2 < pi2 \\ CR & \text{否则} \end{cases}$$



## b6e6rl算法

- 0.5 0.8 0 0.5 1.0 6

- DE/rand/1/bin DE/rand/1/exp 12

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$$q_i = \sum_{j=1}^i p_i$$

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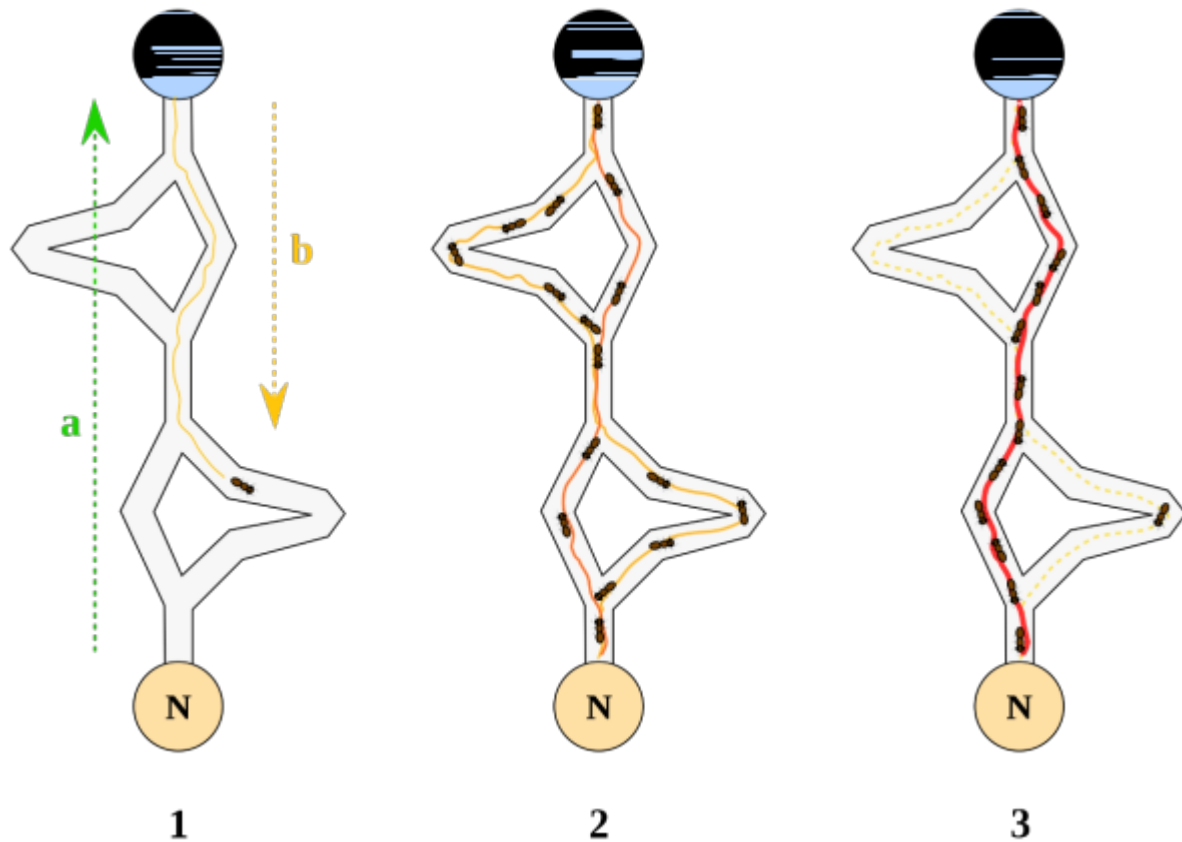


# 差分进化算法差分策略

公式	名称	差分策略	差分表
$DS_1$	DE/rand	DE/rand/bin	$X_i + F * (X_{r1} - X_{r2})$
$DS_2$	DE/rand-to-best	DE/rand-to-best/bin	$X_i + F * (X_{best} - X_i + X_{r1} - X_{r2})$
$DS_3$	DE/best	DE/best/bin	$X_i + F * (X_{best} - X_i)$
$DS_4$	DE/best-to-best	DE/best-to-best/bin	$X_i + F * (X_{best} - X_i + X_{r1} - X_{r2})$
$DS_5$	DE/rand-to-best	DE/rand-to-best/bin	$X_i + F * (X_{best} - X_i + X_{r1} - X_{r2})$
$DS_6$	DE/current-to-best	DE/current-to-best/bin	$X_i + F * (X_{best} - X_i + X_{r1} - X_{r2})$
$DS_7$	DE/current-to-rand	DE/current-to-rand/bin	$X_i + F * (X_{r1} - X_i + X_{r2} - X_{r3})$

# 群 化算法

## 群 化算法原理









# 策略池

- 5
- 1. Classic DE DE/rand/1/bin
- 2. Classic DE DE/rand/1/exp
- 3. Improved JADE IQE
- 4. Improved jDE IQE
- 5. IQE IQE

- JADE jDE NP F CR  $F_i$   
CR<sub>i</sub> NP  $F_i$  CR<sub>i</sub>  $F_i$  CR<sub>i</sub> i  
IQE JADE jDE



# 群 裁剪策略



$$d_{12} = \sqrt{\sum_{k=1}^n (x_{1k} - x_{2k})^2}$$



dmin

$$d_{min} = d_{bw} * \omega$$



# 差分 化中的个体 量 价思想



DE DE





# 群个体 机制



1

2

3



# IQE控制参数 置

$$F'_i = N(\mu, \sigma^2), \quad i = 1, \dots, NP, F'_i \in [0.1, 0.9]$$

~~$$CR'_i = N(\mu, \sigma^2), \quad i = 1, \dots, NP, CR'_i \in [0.1, 0.9]$$~~



# IQE 异策略

- DE/current-to-rand/1/bin  
DE/current-to-best/1/bin

$$v_i = \begin{cases} X_i + F_i * (X_{r1} - X_i) + F_i * (X_{r2} - X_{r3}), & X_i \in \text{优势个体} \\ X_i + F_i * (X_{rs} - X_i) + F_i * (X_{r1} - X_{r2}), & X_i \in \text{劣势个体} \end{cases}$$

- DE/rand-to-rand/1/bin  
DE/rand-to-best/1/bin

$$v_i = \begin{cases} X_{r1} + F_i * (X_{r2} - X_{r1}) + F_i * (X_{r3} - X_{r4}), & X_i \in \text{优势个体} \\ X_{r1} + F_i * (X_{rs} - X_{r1}) + F_i * (X_{r2} - X_{r3}), & X_i \in \text{劣势个体} \end{cases}$$



# 群个体 束放松 理

- - Constraint EC

$$\varepsilon(0) = cv(X_\theta)$$

$$\varepsilon(g) = \begin{cases} \varepsilon(0) \left(1 - \frac{g}{G_c}\right)^{\varepsilon p}, & 0 < g < G_c \\ 0, & g \geq G_c \end{cases} \quad \varepsilon(g)$$







# 群个体 度EDA 理



$$X'_{r,d} = N(X_{r,d}, dp) , d = 1, \dots, D, X'_{r,d} \in [X_d^l, X_d^u]$$



# 典型科研 目

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- Supercontral
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# 文成果

- A mixed integer programming model for gas distribution problem with complex gas applied characteristics,
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# 科研

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- Python .NET