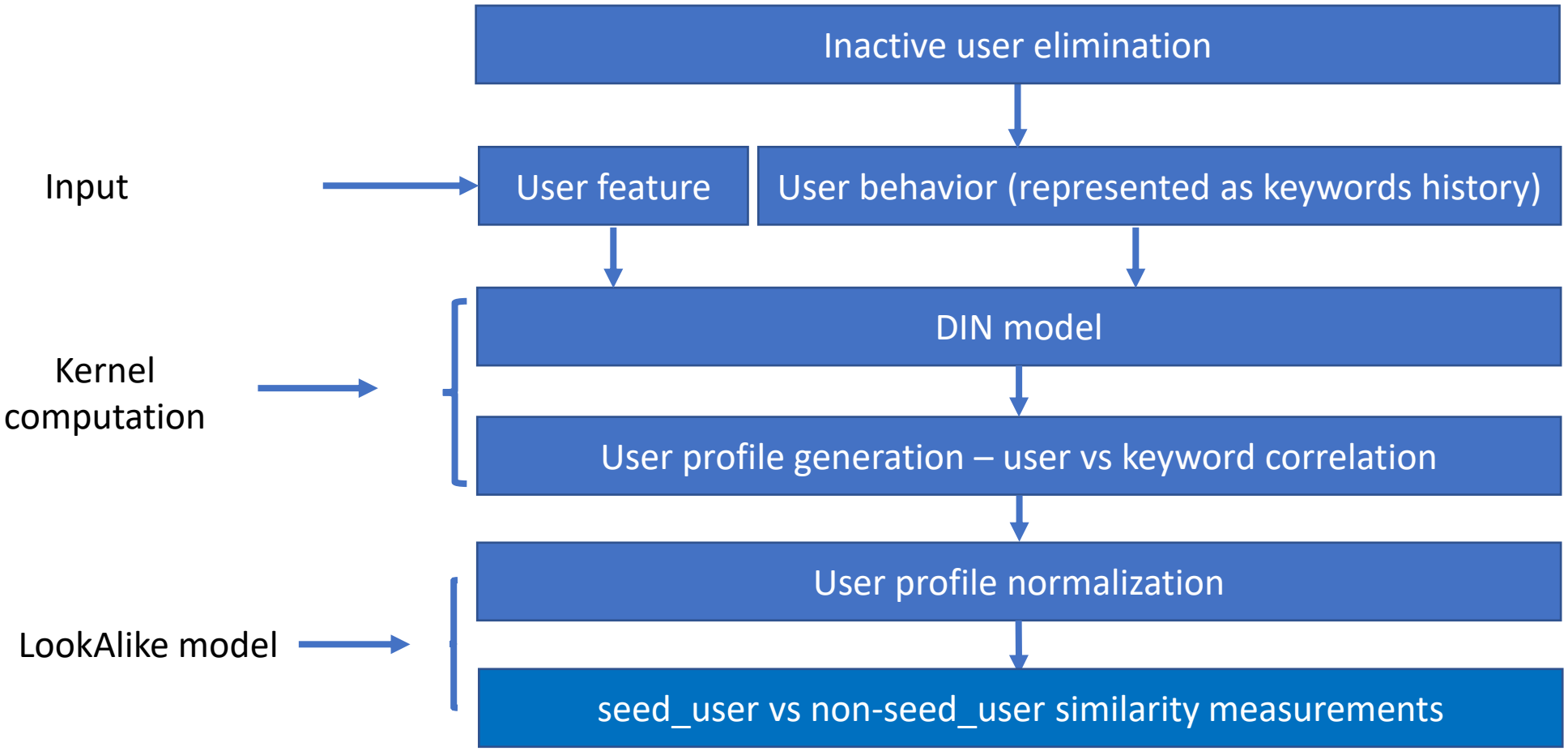
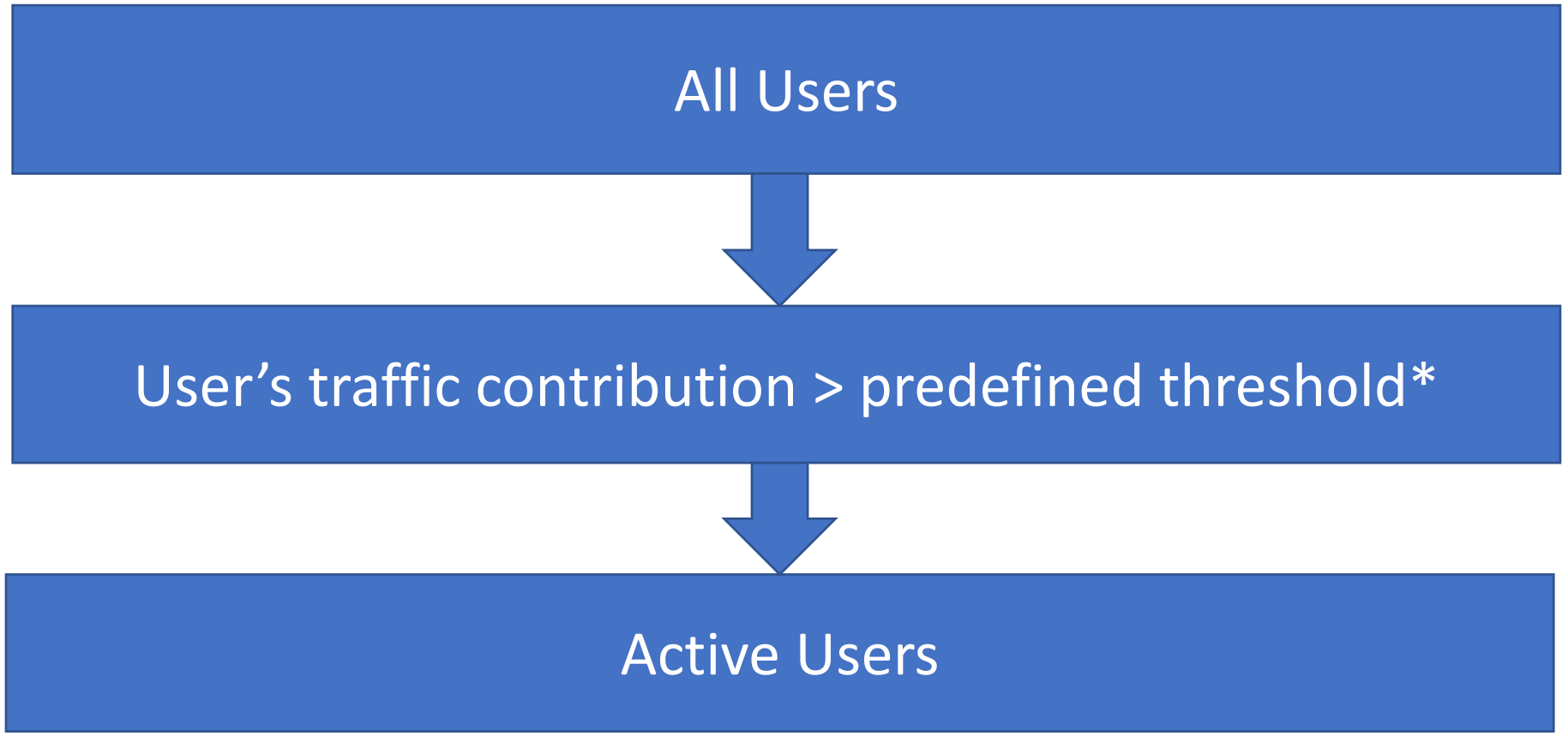
DIN based Look-Alike model

Workflow

- Necessary steps



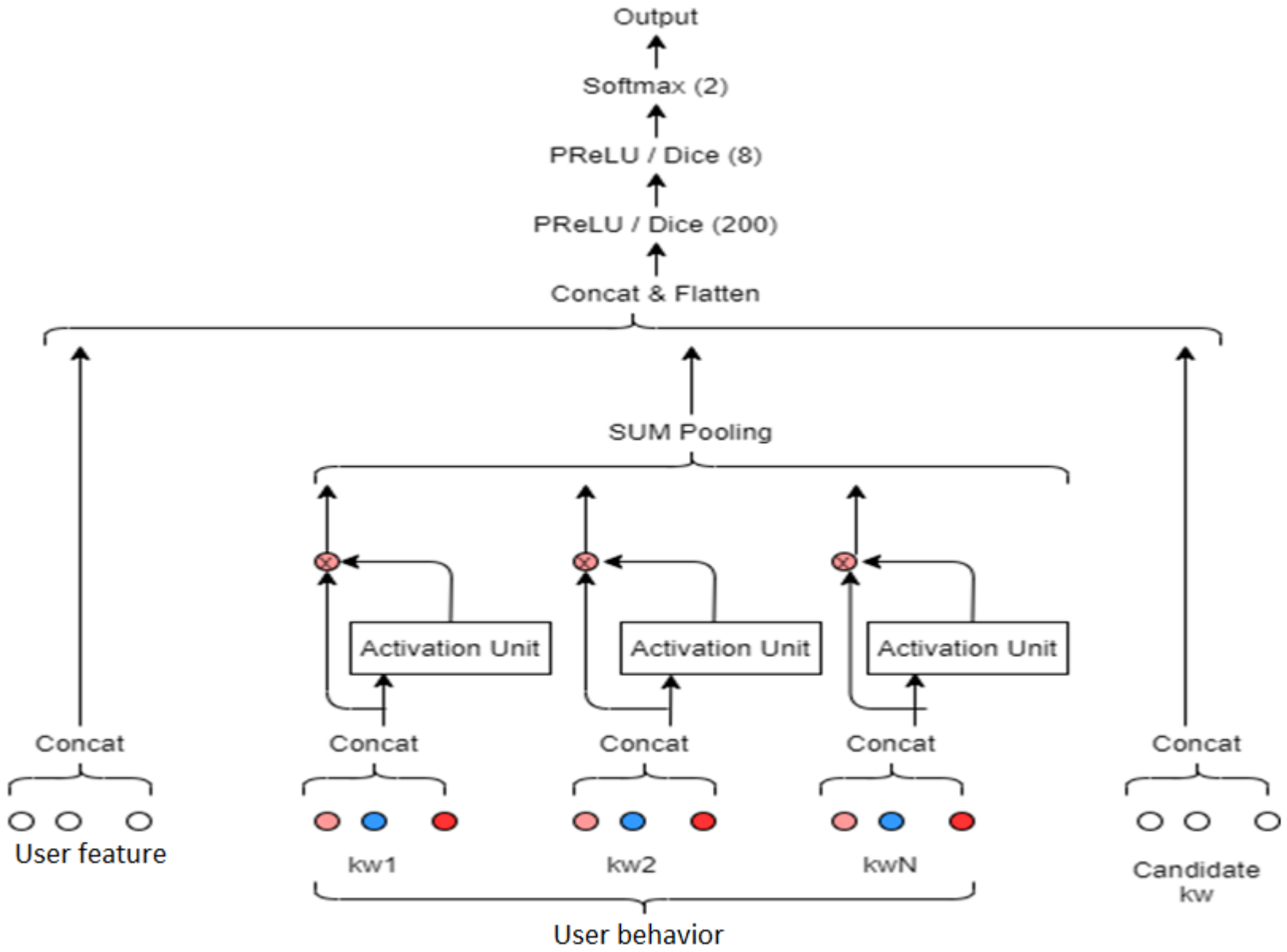
Inactive user elimination (user prescreen)



“Prefined threshold” is defined as a range of normal traffic (with low and high bounds) to eliminate:

* 1. users with consistent low traffic (inactive user, traffic < low bound)
2. users with extremely high traffic for some specific period (robot user, traffic > high bound)

DIN Model


Output
Softmax (2)
PReLU / Dice (8)
PReLU / Dice (200)
Concat & Flatten
SUM Pooling
Activation Unit
Activation Unit
Activation Unit
Concat
Concat
Concat
Concat
User feature
kw1
kw2
kwN
Candidate kw
User behavior

DIN Model Output

– user vs keyword correlation (user profile generation)

	$Keyword_1$	$Keyword_2$	$Keyword_3$	$Keyword_4$	…	$Keyword_m$
$User_1$	$score_{11}$	$score_{12}$	$score_{13}$	$score_{14}$	…	$score_{1m}$
$User_2$	$score_{21}$	$score_{22}$	$score_{23}$	$score_{24}$	…	$score_{2m}$
…	…	…	…	…	…	…
$User_n$	$score_{n1}$	$score_{n2}$	$score_{n3}$	$score_{n4}$	…	$score_{nm}$

DIN Model Output

– user profile normalization

DIN

	$Keyword_1$	$Keyword_2$	$Keyword_3$	$Keyword_4$	…	$Keyword_m$
$User_1$	$score_{11}$	$score_{12}$	$score_{13}$	$score_{14}$	…	$score_{1m}$
$User_2$	$score_{21}$	$score_{22}$	$score_{23}$	$score_{24}$	…	$score_{2m}$
…	…	…	…	…	…	…
$User_n$	$score_{n1}$	$score_{n2}$	$score_{n3}$	$score_{n4}$	…	$score_{nm}$

Normalization constant
C_1
C_2
…
C_n

Score normalization

	$Keyword_1$	$Keyword_2$	$Keyword_3$	$Keyword_4$	…	$Keyword_m$
$User_1$	$norm\_score_{11}$	$norm\_score_{12}$	$norm\_score_{13}$	$norm\_score_{14}$	…	$norm\_score_{1m}$
$User_2$	$norm\_score_{21}$	$norm\_score_{22}$	$norm\_score_{23}$	$norm\_score_{24}$	…	$norm\_score_{2m}$
…	…	…	…	…	…	…
$User_n$	$norm\_\_Score_{n1}$	$norm\_score_{n2}$	$norm\_score_{n3}$	$norm\_score_{n4}$	…	$norm\_score_{nm}$

$$norm\_score_{ij} = \frac{score_{ij}}{C_i}$$

$$C_i = \sqrt{\sum_{j=1}^{m} score_{ij}^2}$$

DIN Model Output

– user similarity measurement

$$\boldsymbol{User_i's\ normalized\ profile}:$$

$$S_i = \{norm\_score_{i1}, \quad norm\_score_{i2}, \quad \dots \quad norm\_score_{im}\}$$

$$\boldsymbol{Cross\ user\ similarity}:$$

$$Similarity(S_i, \quad S_j) = S_i \quad \cdot \quad S_j = \sum_{k=1}^{m} norm\_score_{ik} \times norm\_score_{jk}$$

DIN based Look-Alike model

– seed_user vs non-seed_user similarity measure

	$Seed\_user_1$	$Seed\_user_2$	……	$Seed\_user_m$
$Nonseed\_user_1$	$Similary_{11}$	$Similary_{12}$	……	$Similary_{1m}$
$Nonseed\_user_2$	$Similary_{21}$	$Similary_{22}$	……	$Similary_{2m}$
$Nonseed\_user_3$	$Similary_{31}$	$Similary_{32}$	……	$Similary_{3m}$
$Nonseed\_user_4$	$Similary_{41}$	$Similary_{42}$	……	$Similary_{4m}$
……	……	……	……	……
$Nonseed\_user_n$	$Similary_{n1}$	$Similary_{n2}$	……	$Similary_{nm}$

Parallel computed and only maximum value for each row need to be stored

All Seed Users
$\text{mean}(\underset{i}{\text{top10}}\ similarity_{1i})$
$mean(\underset{i}{\text{top10}}\ similarity_{2i})$
$\text{mean}(\underset{i}{\text{top10}}\ similarity_{3i})$
$\text{mean}(\underset{i}{\text{top10}}\ similarity_{4i})$
……
$\text{mean}(\underset{i}{\text{top10}}\ similarity_{ni})$

sort

$Rank_1$ nonseed_user
$Rank_2$ nonseed_user
$Rank_3$ nonseed_user
$Rank_4$ nonseed_user
…
$Rank_n$ nonseed_user

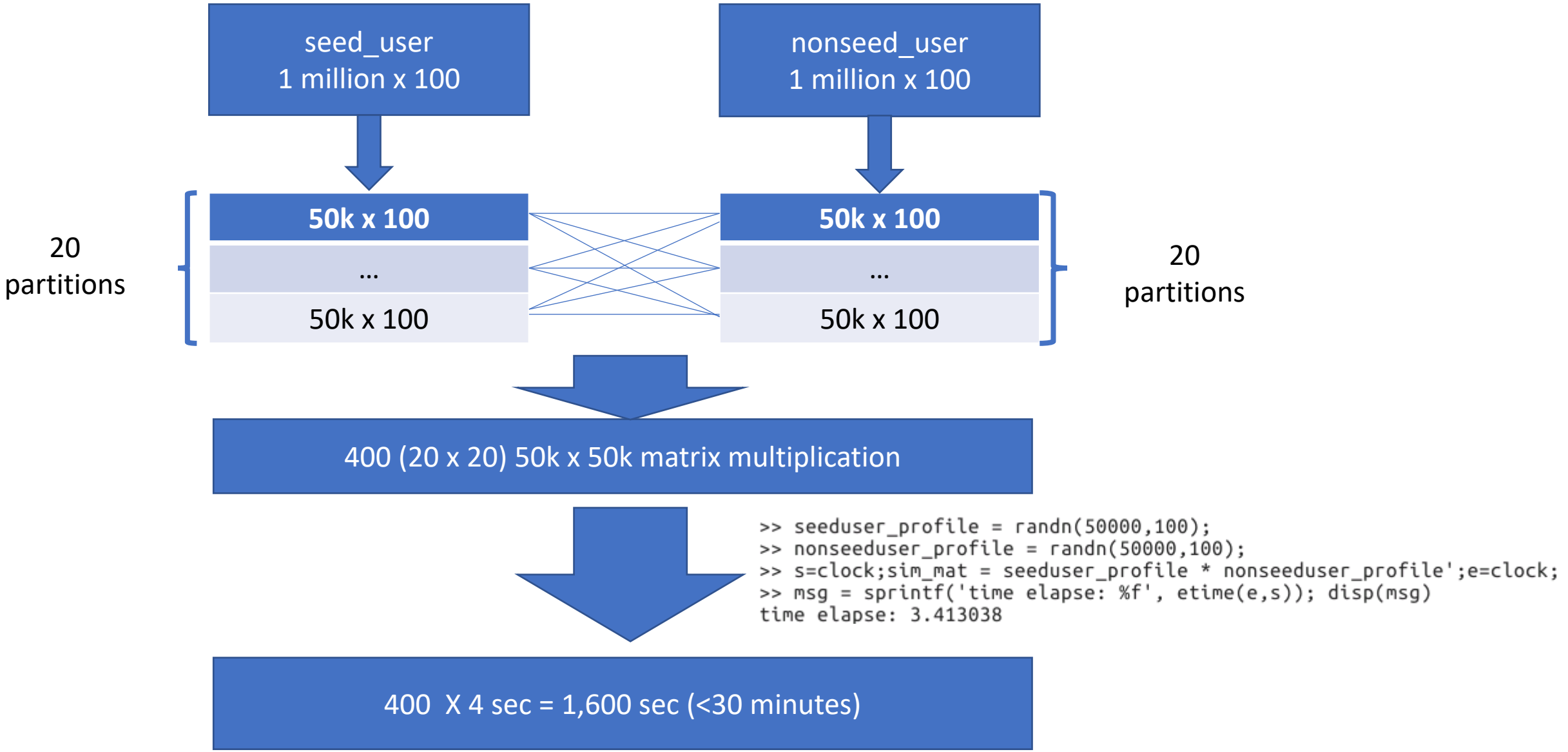
Similarity computation estimation

$$M_{seed} = \begin{bmatrix} norm\_score_{1,1} & \cdots & norm\_score_{1,m} \\ \cdots & \cdots & \cdots \\ norm\_score_{n_{seed},1} & \cdots & norm\_score_{n_{seed},m} \end{bmatrix}$$

$$M_{nonseed} = \begin{bmatrix} norm\_score_{1,1} & \cdots & norm\_score_{1,m} \\ \cdots & \cdots & \cdots \\ norm\_score_{n_{nonseed},1} & \cdots & norm\_score_{n_{nonseed},m} \end{bmatrix}$$

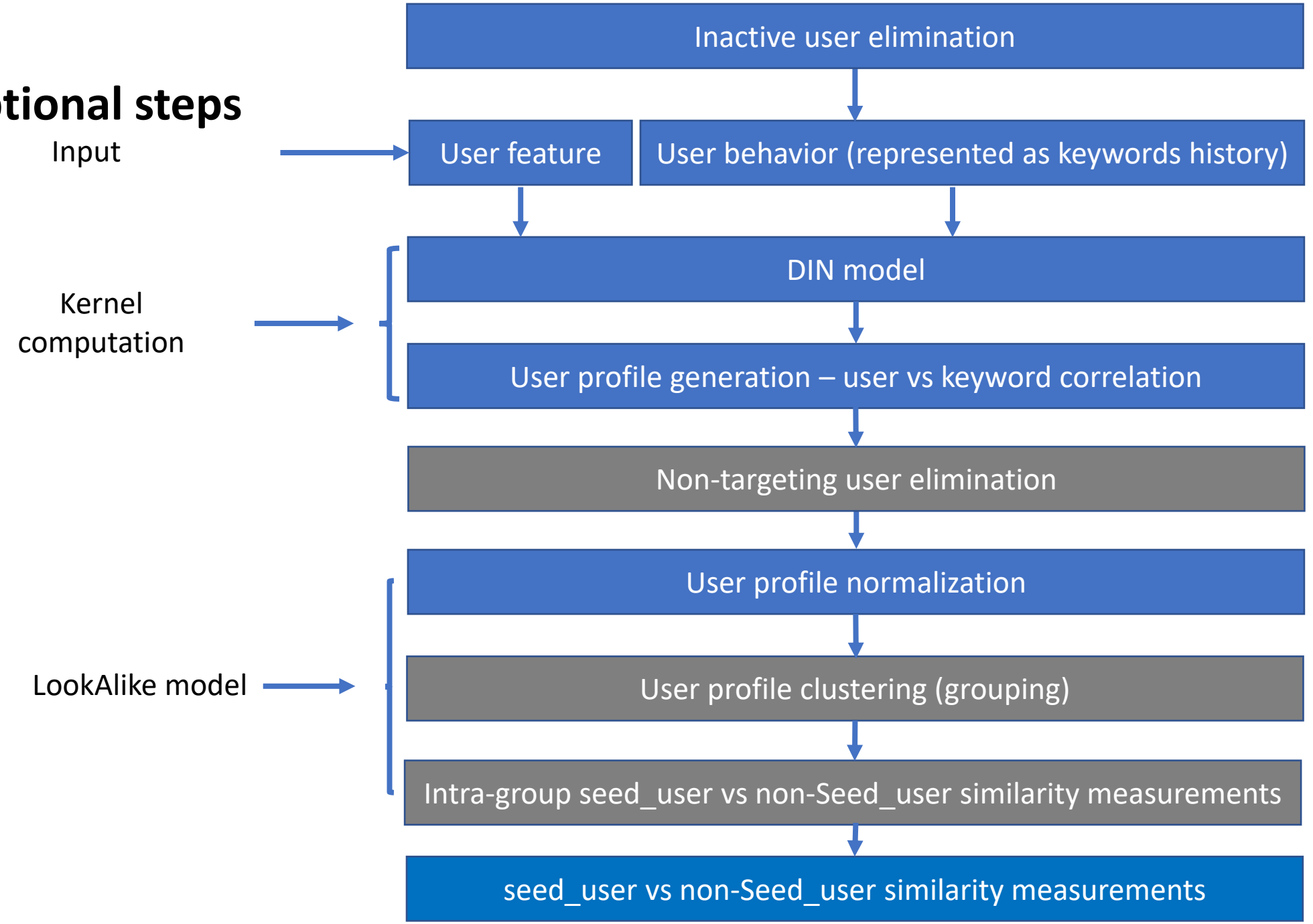
$$M_{similarity} = M_{seed} \quad \times \quad M_{nonseed}^{T}$$

Similarity computation estimation



Workflow

- Necessary + optional steps



DIN Model Output

– non-targeting user elimination

	$Keyword_1$	$Keyword_2$	$Keyword_3$	$Keyword_4$	…	$Keyword_m$
$User_1$	$score_{11}$	$score_{12}$	$score_{13}$	$score_{14}$	…	$score_{1m}$
$User_2$	$score_{21}$	$score_{22}$	$score_{23}$	$score_{24}$	…	$score_{2m}$
…	…	…	…	…	…	…
$User_n$	$score_{n1}$	$score_{n2}$	$score_{n3}$	$score_{n4}$	…	$score_{nm}$

DIN

$$User_i's\ profile:\ S_i = \{score_{i1},\ score_{i2},\ \dots\ score_{im}\}$$

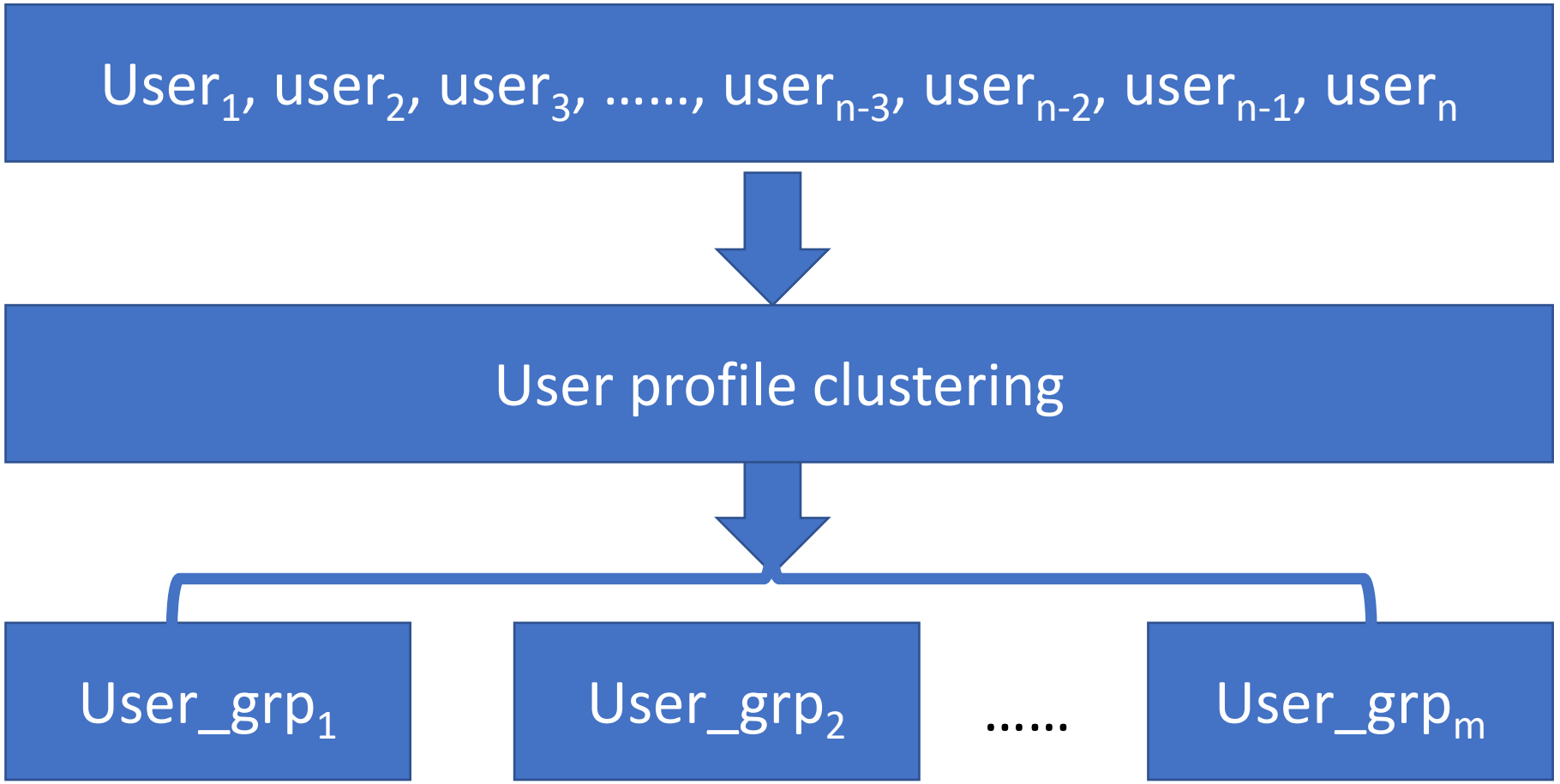
Eliminate $max(S_i) = \max_j score_{ij} < prefined\ threshold$

$targeting\ users$

* The purpose is to eliminate users that have no interest of any keywords (ineffective traffic)

DIN Model Output

– user clustering



DIN based Look-Alike model

– group-wise seed_user vs non-seed_user similarity measure (active user only)

	Seed user in grp_1	Seed user in grp_2	……	Seed user in grp_m
Nonseed user in grp_1	similarity $matrix_{11}$	0	……	0
Nonseed user in grp_2	0	similarity $matrix_{22}$	……	0
……	……	……	……	0
Nonseed user in grp_m	0	0	……	similarity $matrix_{mm}$

DIN based Look-Alike model

– within group seed_user vs non-seed_user similarity measure

Similarity matrix$_{ji}$

	$Seed\_user_{grpi,1}$	$Seed\_user_{grpi,2}$	……	$Seed\_user_{grpi,m}$
$Nonseed\_user_{grpi,1}$	$Similary_{11}$	$Similary_{12}$	……	$Similary_{1m}$
$Nonseed\_userg_{rpi,2}$	$Similary_{21}$	$Similary_{22}$	……	$Similary_{2m}$
$Nonseed\_user_{grpi,3}$	$Similary_{31}$	$Similary_{32}$	……	$Similary_{3m}$
$Nonseed\_user_{grpi,4}$	$Similary_{41}$	$Similary_{42}$	……	$Similary_{4m}$
……	……	……	……	……
$Nonseed\_user_{grpi,n}$	$Similary_{n1}$	$Similary_{n2}$	……	$Similary_{nm}$

Parallel computed and top 10 values for each row need to be stored

All Seed Users in grpi
$\underset{i}{\text{mean}}(\text{top10}\ similarity_{1i})$
$\underset{i}{mean}(\text{top10}\ similarity_{2i})$
$\underset{i}{\text{mean}}(\text{top10}\ similarity_{3i})$
$\underset{i}{\text{mean}}(\text{top10}\ similarity_{4i})$
……
$\underset{i}{\text{mean}}(\text{top10}\ similarity_{ni})$

sort

$Rank_1$ nonseed_user
$Rank_2$ nonseed_user
$Rank_3$ nonseed_user
$Rank_4$ nonseed_user
…
$Rank_n$ nonseed_user