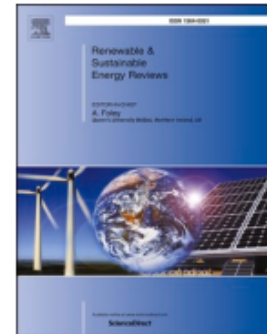




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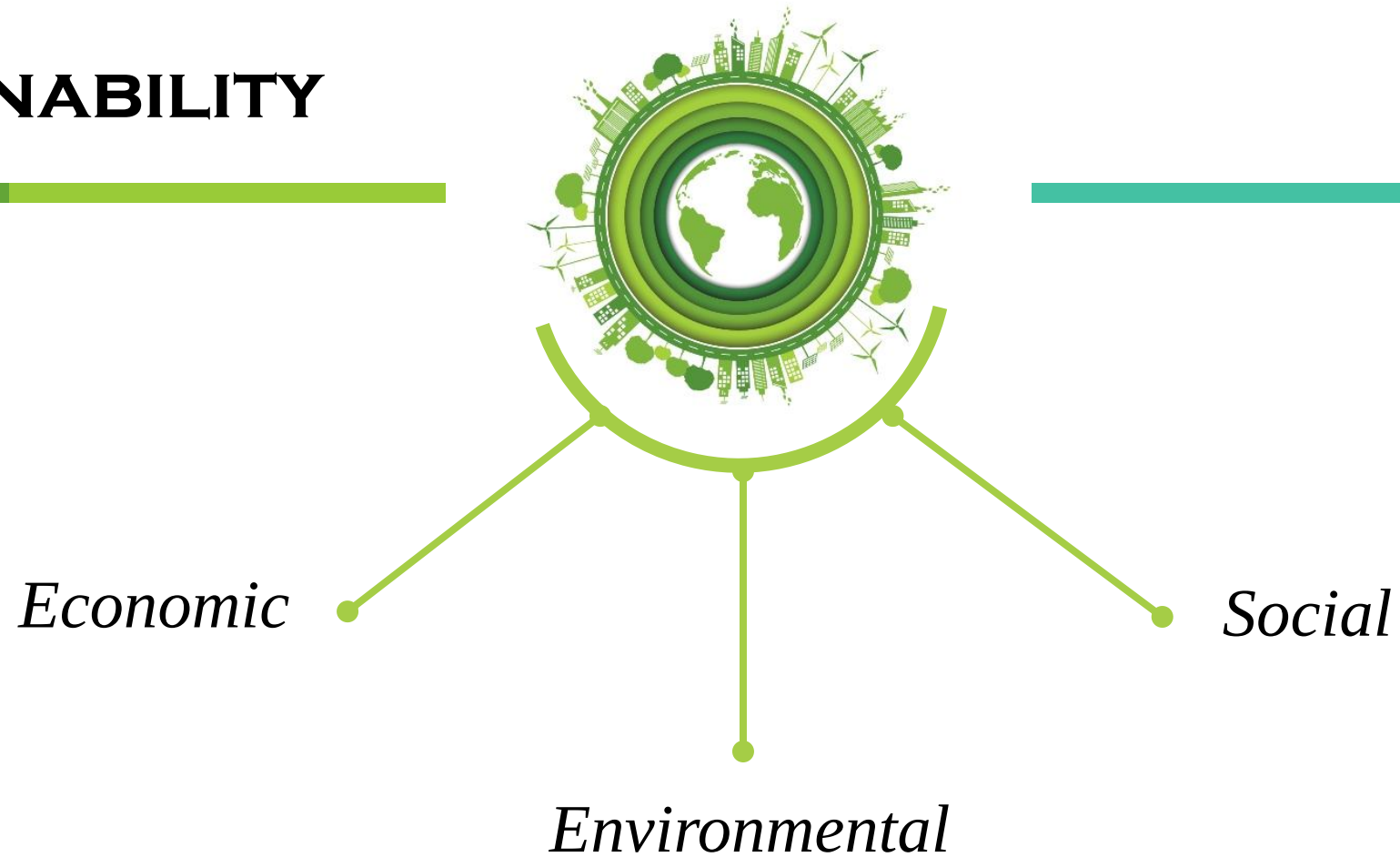
Designing a sustainable closed-loop supply chain network for walnut industry

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SUSTAINABILITY



CLOSED-LOOP

SUPPLY CHAIN



INTRODUCTION

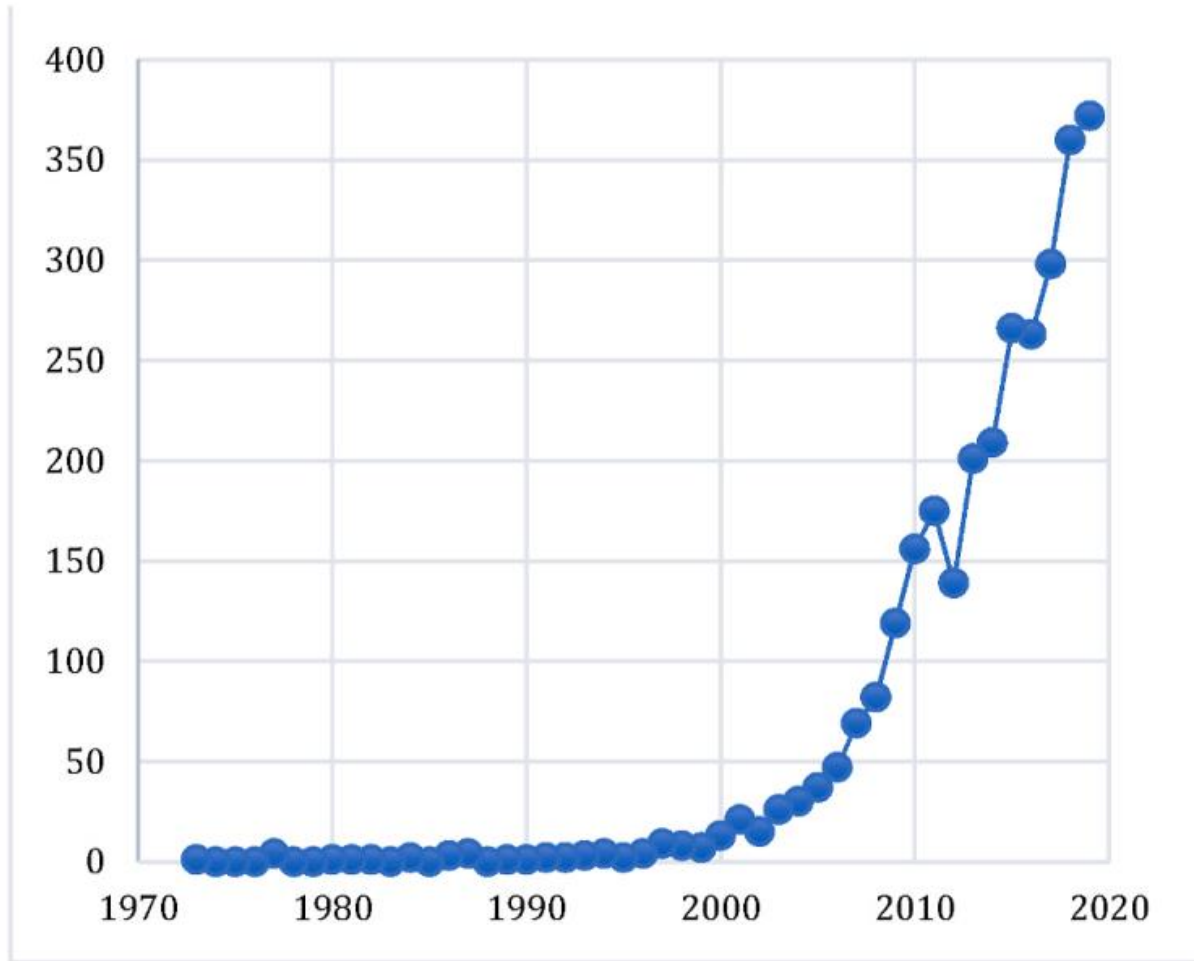


ASC

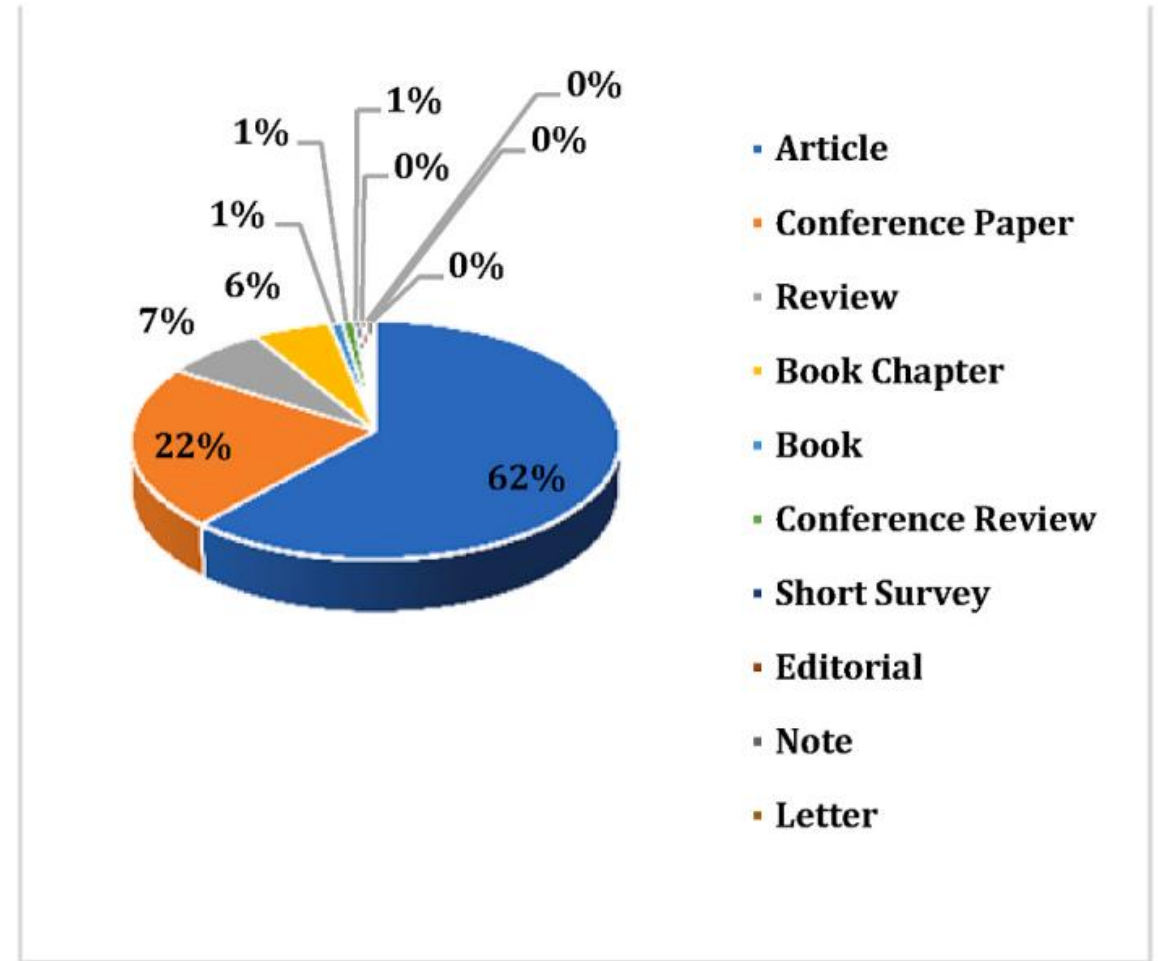
Usages & Benefits

Nuts/Walnut

FIG 1. Trend of publications in the field of agriculture supply chain: (a) number (b) publication category.

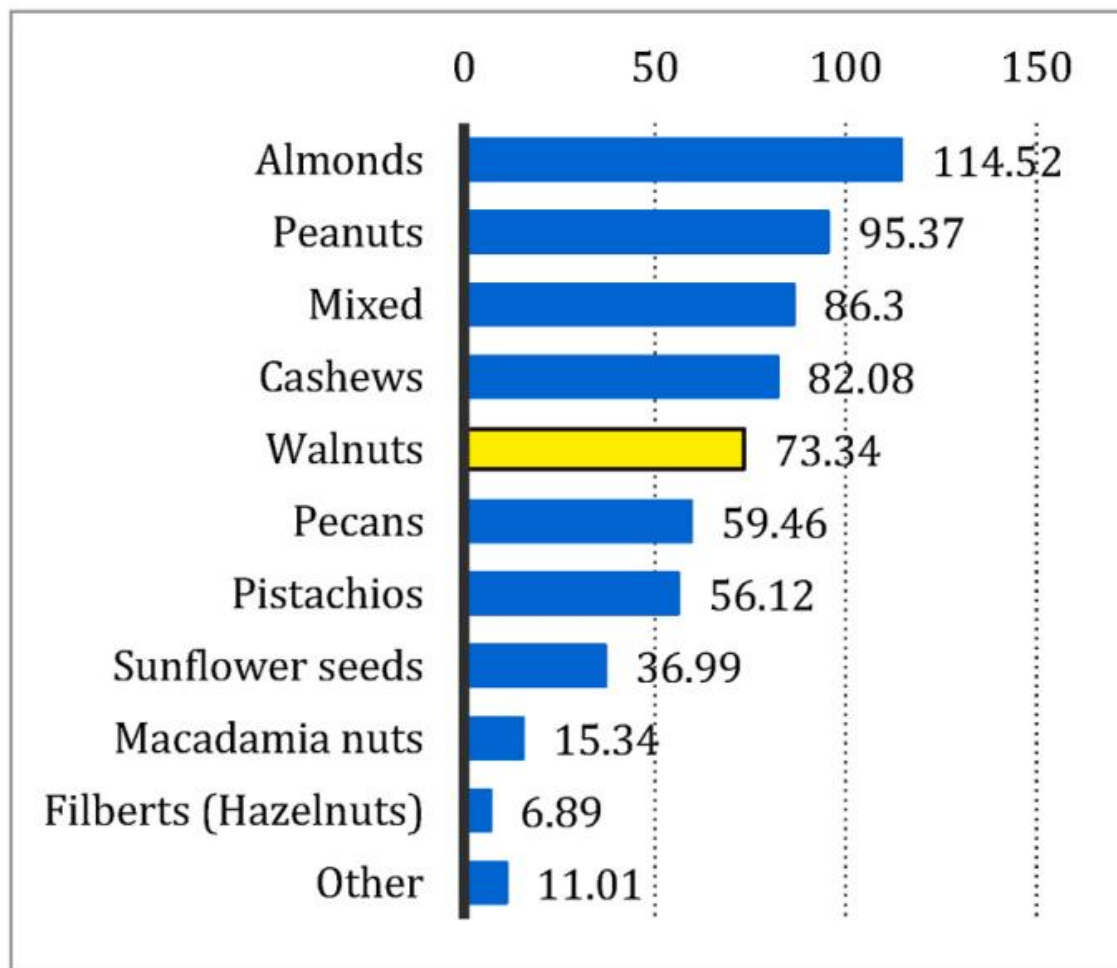


(a)

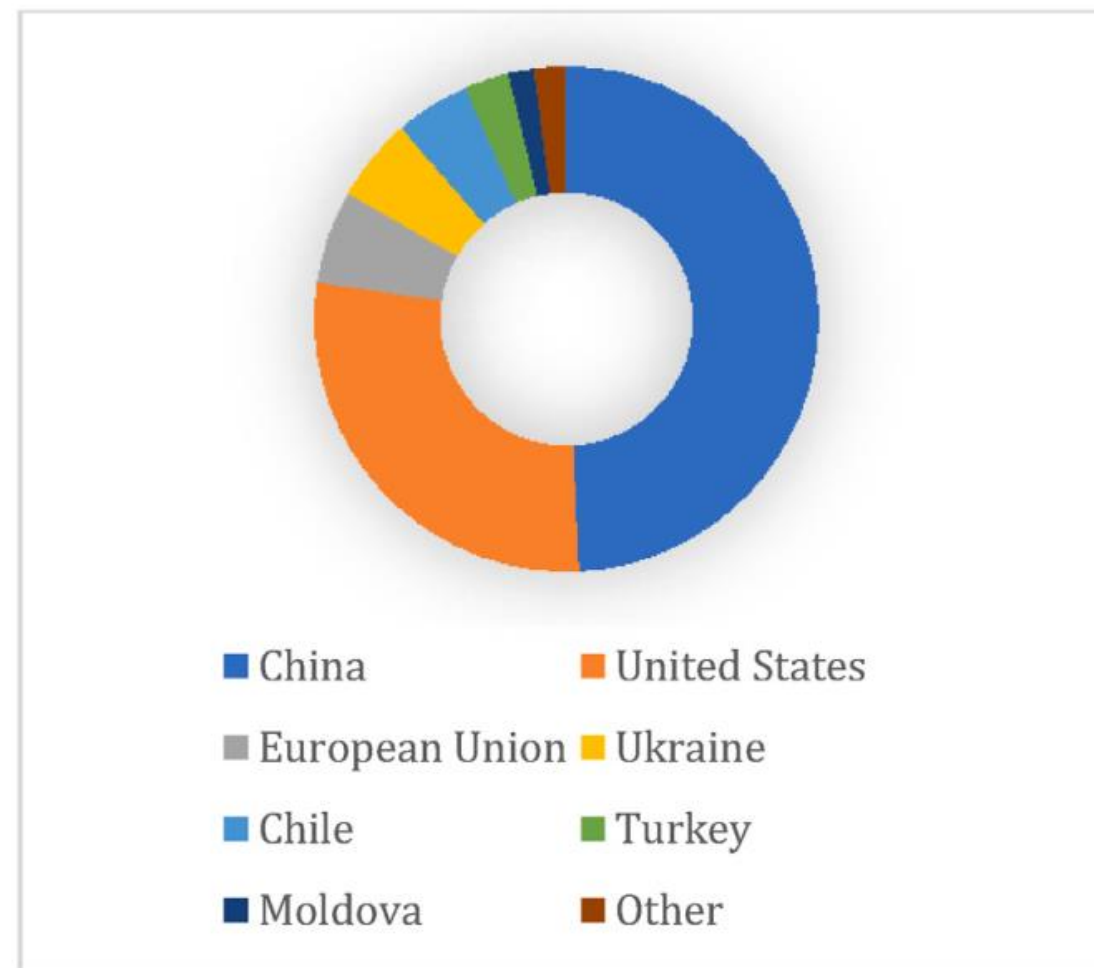


(b)

FIG 2. Walnut production share (a) US consumption (b) global production share.



(a)



(b)

FIG 5. Walnut constituents and their usage.

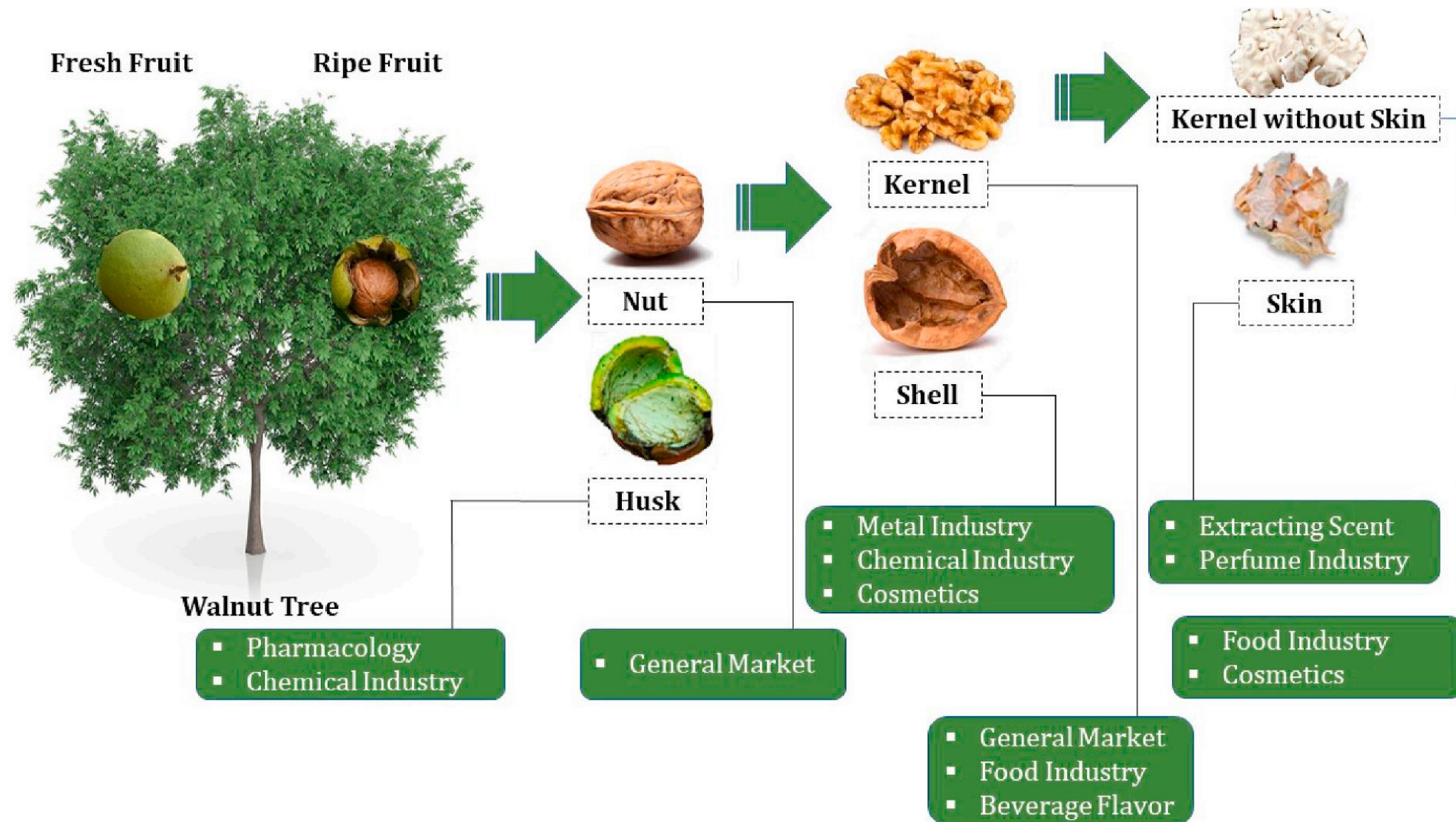


FIG 6. Walnut constituents and their usage.

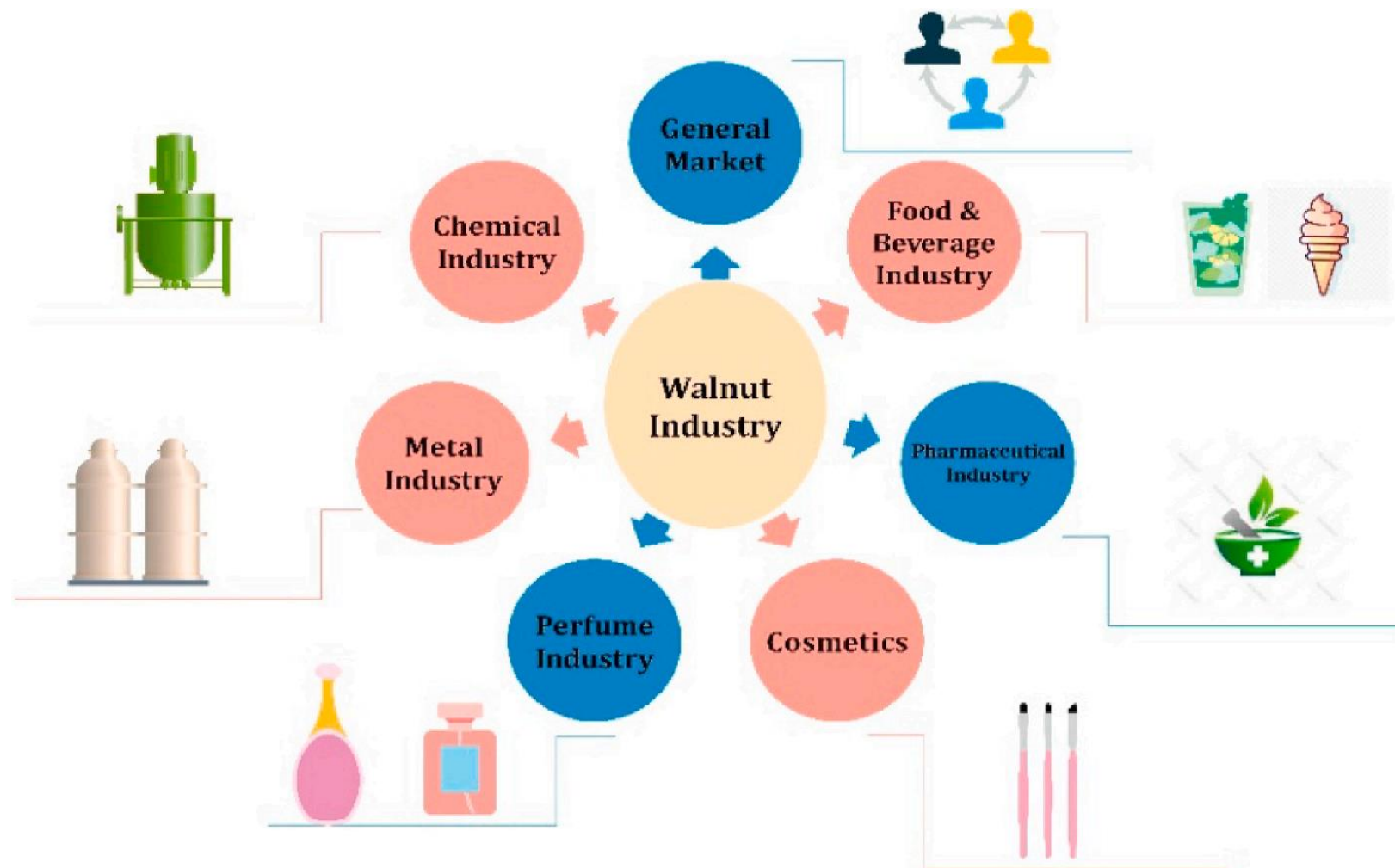
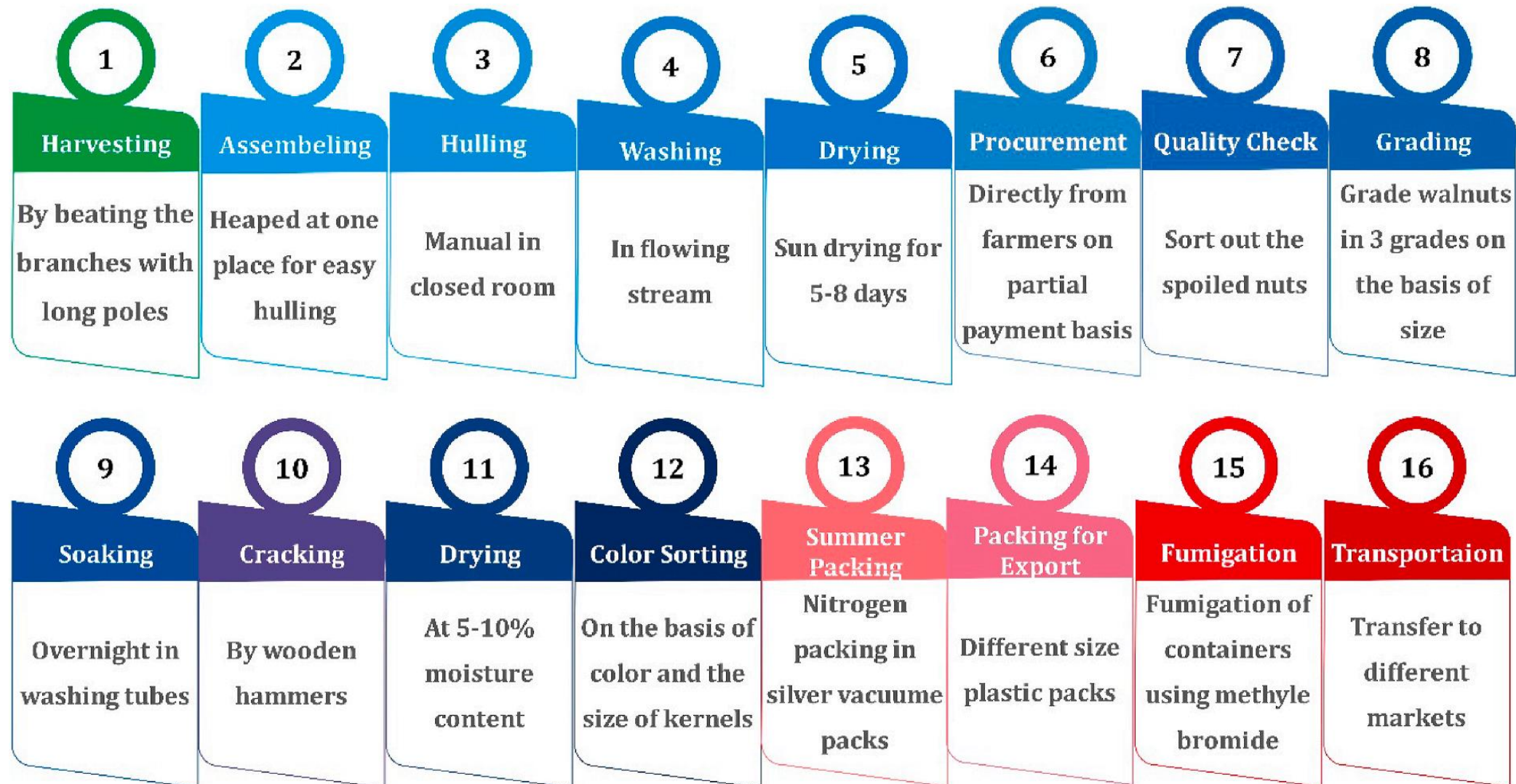


FIG 7. The main process for walnut value chain



MATHEMATICAL

FORMULATION

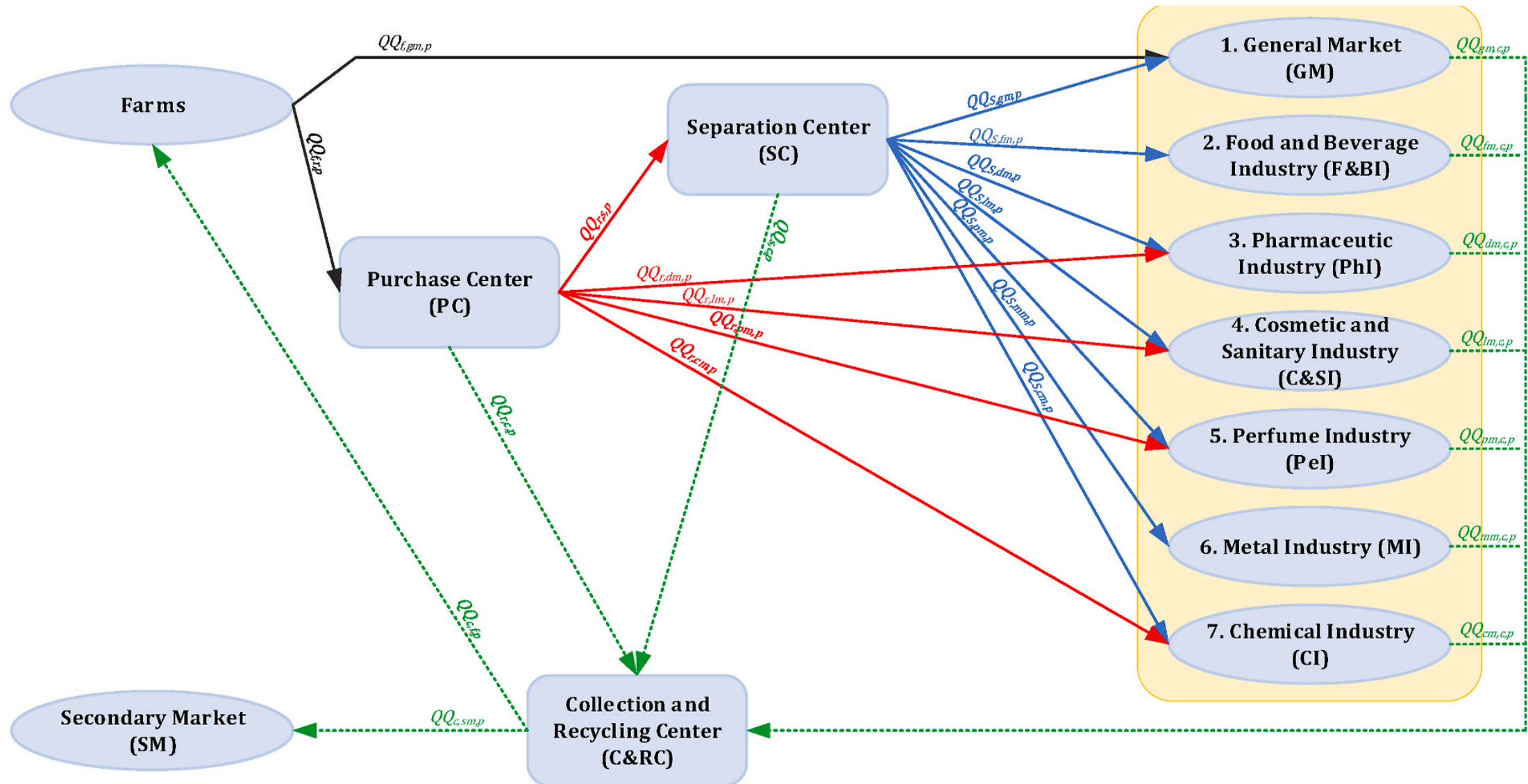


Network

Parameters

Model

FIG 9. Structure of the walnut supply chain network.



SOLUTION

APPROACH



Meta-Heuristics

Hybrid Meta-Heuristics

SA - SEO - KA

HKSA - HKSEO

ENCODING



DECODING

Section 5					
c		$sm+f$			
0.81	0.04	0.33	0.62	0.26	0.25
1	6	3	2	4	5

FIG 13. The pseudo-code of the SA algorithm.

Solution: initial random solution generation

T: initial temperature

Value: solution evaluation

While ceasing condition not confirmed *Do*

for B= 1: Maximum sub iteration *Do*

 New solution: neighborhood solution generation

 New Value: New solution evaluation

if new value < value

 solution=new solution

 value=new value

else

$\Delta = \text{new value} - \text{value}$

 acceptation of new solution probability = $e^{\left(\frac{\Delta}{T}\right)}$

end if

end for

 reduce T

end while

FIG 13. The pseudo-code of the SEO algorithm.

~

```
Initialize an attacker and defender
It=0;
while It <Maxit
    Do training and retraining;
    Num_attack=0;
    while Num_attack <Max_attack
        Spot an attack;
        Check the boundary;
        Respond to attack,
        if the Objective Function (OF) of the defender
        is lower than the attacker
            Exchange the defender and attacker position;
        end if
        Num_attack=Num_attack+1;
    end while
    Create a new solution as a defender,
    It=It+1;
end while
Return the attacker
```

```

Initialize feasible Keshtels population randomly
Calculate the fitness
Sort Keshtels in three types:  $N_1$  as the best,  $N_2$ , and  $N_3$  as the
worst solutions
  for i = 1 to MaxIt
    for each  $N_1$ 
      Select a chromosome  $a_i$  from  $N_1$  population
      Find the nearest neighbor to  $a_i$  in the populations
      Determine the new solution  $b_i$  by finding distance
      between this best Keshtel  $a_i$  and between neighbors.
      Calculate the cost function of the new solution
      If new solution  $a_i$  is better than the  $b_i$ , then  $a_i = b_i$ 
      else
        Considering half the radius generates two solutions
         $b_{i1}$  and  $b_{i2}$ 
        end if
        If new solution  $b_{i1}$  or  $b_{i2}$  is better than  $a_i$  then replace
        it with  $a_i$ 
        else
          use the final generated solution as  $a_i$ 
        end if
      end for
    for each  $N_2$ 
      selects two chromosomes randomly
      move the chromosome of the population  $N_2$ 
      according to two selected chromosomes
    end for
    Generates  $N_3$ 
    chromosomes randomly
    merge population  $N_1$ ,  $N_2$  and  $N_3$ 
    sort the population
  end for

```

FIG 15. The pseudo-code of the KA algorithm.

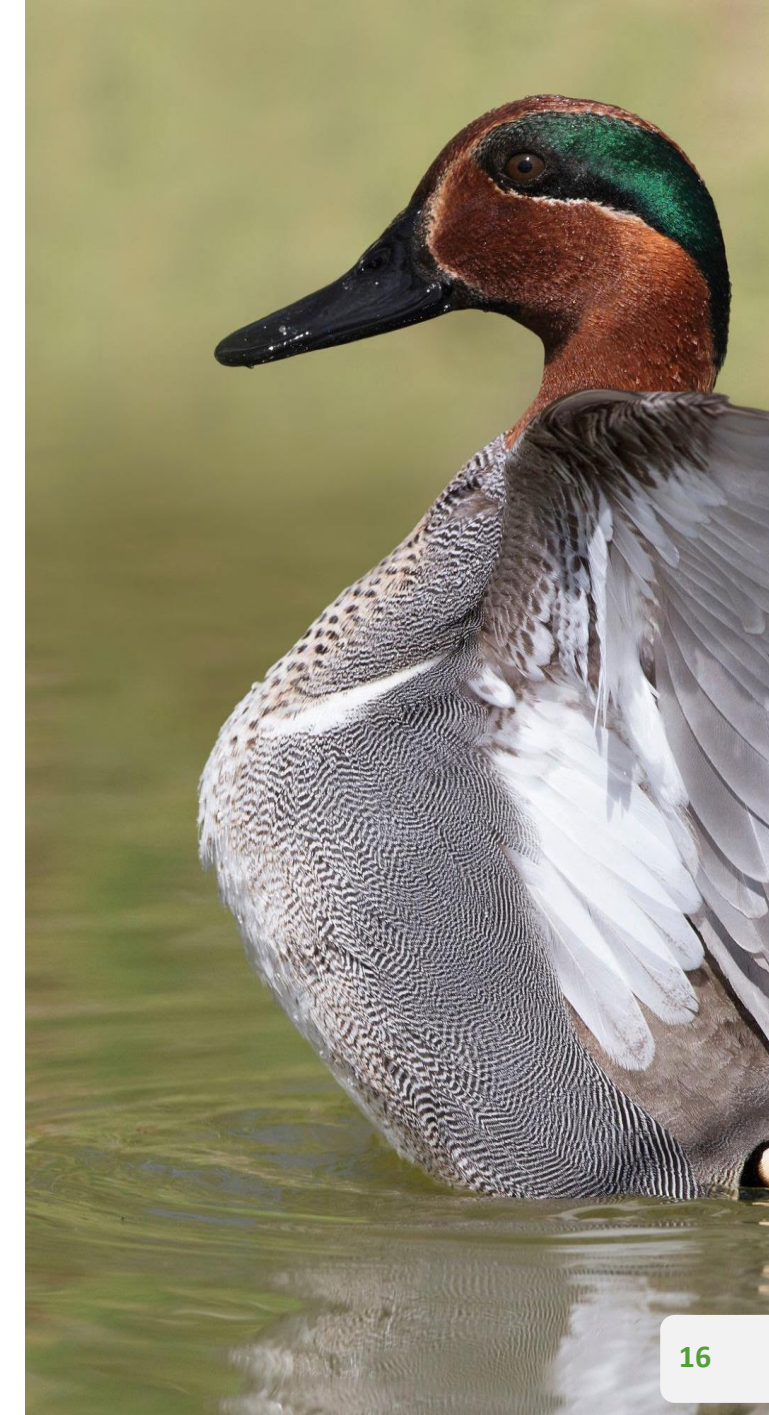


FIG 16-17. The pseudo-code of H-KASA & H-KASEO.

```
Initialize Keshtels population
Calculate the fitness and sort them in three types:  $N_1$ ,  $N_2$ , and  $N_3$ 
 $X^*$ =the best solution.
while ( $t < \text{maximum number of iteration}$ )
    for each  $N_1$ 
        Do SA
        if the fitness of this new position is better than prior
            Update this lucky Keshtel.
        endif
    endfor
    for each  $N_2$ 
        Move the Keshtel between the two Keshtels.
    endfor
    for each  $N_3$ 
        Generate new Keshtels as new random solutions.
    endfor
    Update T;
    Merge the  $N_1$ ,  $N_2$ , and  $N_3$ 
    Sort the Keshtels and form  $N_1$ ,  $N_2$ , and  $N_3$  for next iteration
    Update the  $X^*$  if there is better solution.
     $t=t+1$ ;
end while
return  $X^*$ 
```

```
Initialize Keshtels population
Calculate the fitness and sort them in three types:  $N_1$ ,  $N_2$ , and  $N_3$ 
 $X^*$ =the best solution.
while ( $t < \text{maximum number of iteration}$ )
    for each  $N_1$ 
        Do SEO
        if the fitness of this new position is better than prior
            Update this lucky Keshtel.
        endif
    endfor
    for each  $N_2$ 
        Move the Keshtel between the two Keshtels.
    endfor
    for each  $N_3$ 
        Generate new Keshtels as new random solutions.
    endfor
    Update T;
    Merge the  $N_1$ ,  $N_2$ , and  $N_3$ 
    Sort the Keshtels and form  $N_1$ ,  $N_2$ , and  $N_3$  for next iteration
    Update the  $X^*$  if there is better solution.
     $t=t+1$ ;
end while
return  $X^*$ 
```

EXPERIMENTAL



RESULTS

Table 7
Problem classification.

Classification	Instance	Problem size (f, r, s, gm, fm, dm, lm, pm, mm, cm, c, sm, P, T[month])
Small	SP1	(3,3,3,1,1,1,1,1,1,1,1,6,1)
	SP2	(10,10,10,3,3,3,3,3,3,5,3,6,3)
	SP3	(15,15,15,6,6,6,6,6,6,10,6,6,6)
Medium	MP4	(30,30,30,6,6,6,6,6,6,15,6,6,9)
	MP5	(40,45,45,8,8,8,8,8,8,20,8,6,11)
	MP6	(50,60,60,12,12,9,9,9,9,25,12,6,12)
	MP7	(50,60,60,15,15,10,10,10,10,25,20,6,12)
Large	LP8	(100,100,100,15,15,12,12,12,12,15,40,30,6,12)
	LP9	(150,150,150,25,25,20,20,20,20,20,60,30,6,12)
	LP10	(200,200,200,50,60,30,30,30,30,80,40,6,12)

$$RPD = \frac{|Alg_{sol} - Min_{sol}|}{Min_{sol}}$$

EXPERIMENTAL

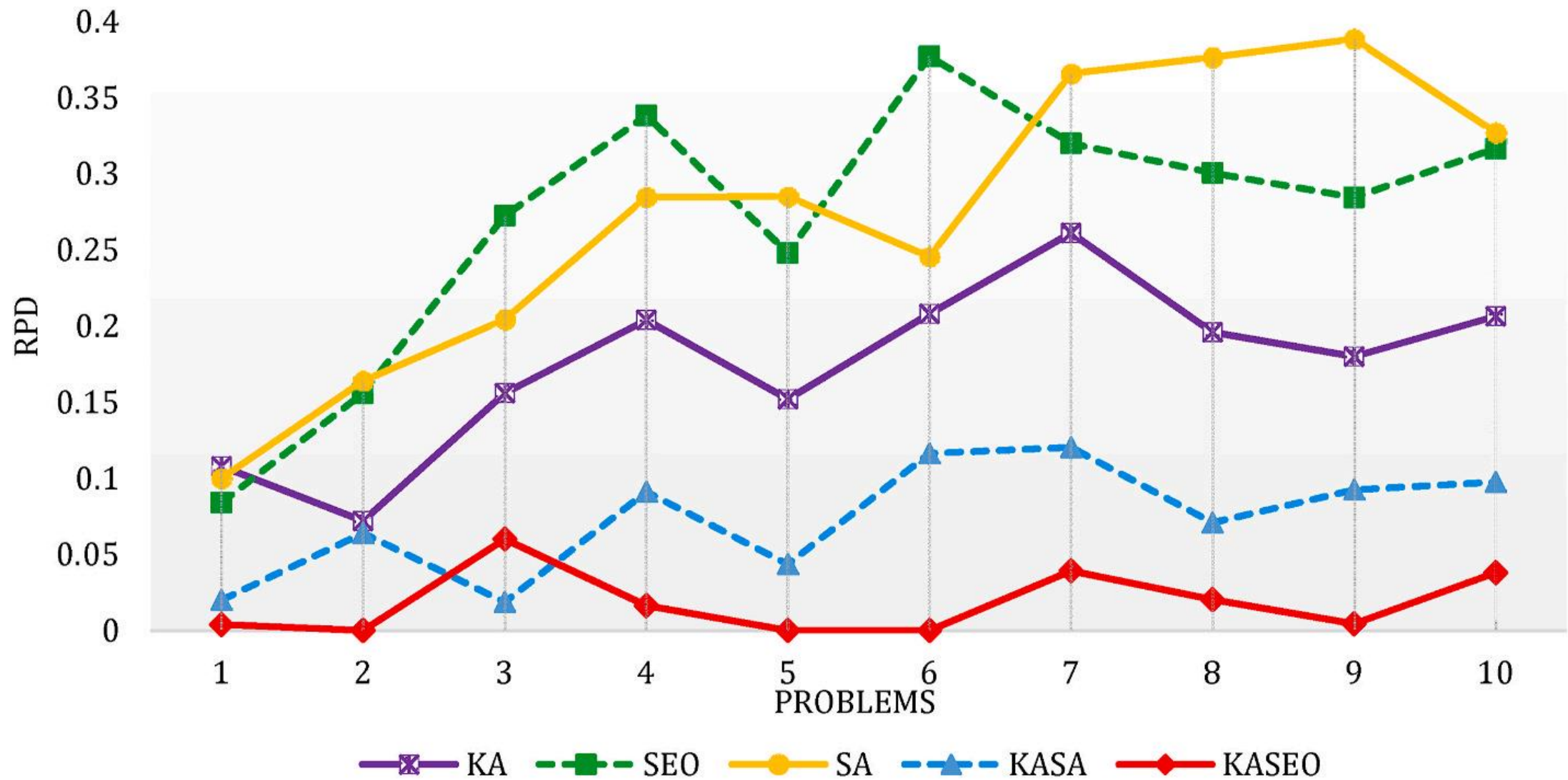


RESULTS

Table 11
The output results.

Objective	SP ₁	SP ₂	SP ₃	MP ₄	MP ₅	MP ₆	MP ₇	LP ₈	LP ₉	LP ₁₀
Gams	704.2	1016	2572	5037	7625	–	–	–	–	–
SA	785	1072	2843	5309	7460	9475	10910	13470	16238	19604
SEO	740	1056	2741	5307	7410	9474	10863	13243	16264	19550
KA	730	1048	2711	5294	7350	9268	10842	13095	16284	19490
KASA	718	1043	2638	5220	7310	9252	10633	13053	16199	19400
KASEO	705	1016	2618	5194	7270	9121	10529	13042	16151	19320

FIG 18. The mean RPD of each algorithm.



Q & A

