Journal of Mechanical Engineering Science and Technology Vol. 4, No. 1, July 2020, pp. 14-27 ISSN 2580-0817 14

Thermal Design Optimization of Shell-and-Tube Heat Exchanger Liquid to Liquid to Minimize Cost using Combination Bell-Delaware Method and Genetic Algorithm

Reza Setiawan^{1*}, František Hrdlička², Prihadi Setyo Darmanto³, Vera Pangni Fahriani⁴, Suciani Rahma Pertiwi⁵

 ¹Department of Mechanical Engineering, Universitas Singaperbangsa Karawang, Jl. H. S. Ronggowaluyo, Karawang, 41360, Indonesia
 ²Department of Energy Engineering, Czech Technical University in Prague, Technicka 4, Praha, 16607, Czech Republic
 ³Department of Mechanical Engineering, Institut Teknologi Bandung, Jl. Ganesha 10, Bandung, 40132, Indonesia
 ⁴Department of Chemical Engineering, Universitas Singaperbangsa Karawang, Jl. H. S. Ronggowaluyo, Karawang, 41360, Indonesia
 ⁵Department of Electrical Engineering, Universitas Singaperbangsa Karawang, Jl. H. S. Ronggowaluyo, Karawang, 41360, Indonesia
 ⁵Department of Electrical Engineering, Universitas Singaperbangsa Karawang, Jl. H. S. Ronggowaluyo, Karawang, 41360, Indonesia

ABSTRACT

Shell-and-tube heat exchanger is designed to satisfy certain requirements such as heat transfer capability, allowable pressure drop and limitation of size. Beside such requirements, it is important to consider economical point of view to get the lowest total cost. In this study, computational program and optimization for thermal design shell-and-tube heat exchanger were built for liquid to liquid with no phase change process in four variables design parameters using Bell-Delaware method. The design variables were tube size, tube length, baffle cut to shell inside diameter ratio and central baffle spacing to shell inside diameter ratio. The genetic algorithm was used as optimization method to get lower solution for economical point of view. The results from two study cases show that the genetic algorithm got lower total cost from the original design. The total cost decreased 28.83% in first study case and 52.56% in second study case from the original design.

Copyright © 2020. Journal of Mechanical Engineering Science and Technology

All rights reserved

Keywords: Bell-Delaware, genetic algorithm, minimizing cost, optimization, shell-and-tube heat exchanger

I. Introduction

The heat exchanger is an important equipment in the industrial process. One of their types has widely used in industrial energy, petroleum industry and chemical process industry. It is designed based on their characteristics and conditions of fluids, and some design is possible to appear similarly for a particular purpose. In such design, heat transfer capability and pressure drop may similar although they have different dimension and arrangement construction. Because it is possible to get many variants design shell and tube through differences of construction, It is better to have design considering economical point of view. The design should consider total cost from investment and operational cost. The cost of investment is defined as a cost for manufacturing of shell and tube and cost of operation is defined as a cost which is needed along the operational process, and actually, it



is a cost for pumping power. The design with low total cost will have a significant impact to expense for producers and users because it commonly is used for a long time or around ten years. In the other hand, computational processes are developed rapidly, and one of them is global random search methods. The uniqueness of this method can find a global optimum point in all problems of optimization. Genetic algorithm is adapted from natural processes. The genetic algorithm mimics from natural process to transmit heredity characteristics from a parent to an offspring by genes in chromosomes. In addition, the method can be used easier to be implemented for the iterative calculation of optimization because some supporting mathematical software can help to build algorithms. The calculation process combined with the best method and supporting software, can solve the design, which has the cheapest cost. Their components are different depending on type shell and tube particularly. But main components are shell, tubes, front-end head, rear-end head and baffles as shown in Fig. 1.

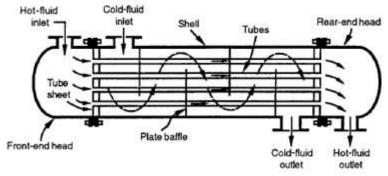


Fig. 1. Main parts of shell-and-tube heat exchanger [1]

There are many standards of shell-and-tube heat exchanger. Tubular Exchanger Manufacture Association (TEMA) standards are widely recognized in many producers and consumers around the world to be used as a standard. TEMA standards are made by engineering principles, researchers and experiences in process design, manufacture and installation to assist designer, engineers and users to work. TEMA standards cover fabrication tolerances and performance information, installation, operation and maintenance. mechanical standards, vibration standards, thermal relations and recommended good practices [1].

Some researchers used algorithms to the optimization of a heat exchanger. Extensive advanced optimization techniques were applied. Trial and error were conducted in various parameter design and operation, and it is very useful to the industry [2]. The study of optimization using particle swarm algorithm in double pipe was employed micro-finned tubes using number of micro-fins from 10 to 60, micro-fin height varying from 0.0 to 0.5 mm and the micro-fin helix angle between 5 and 30° [3]. Grey wolf optimization algorithm reduced total cost using relatively low computing time [4]. Elitist-Jaya algorithm in first case reduced 32.855% and in the second case reduced 5.21% from simulations result compared to original design for continuous parameter optimization [5]. In the current investigation, Multi-Objective optimization of Bees Algorithm Hybrid and Particle Swarm purposed to acquire the maximum effectiveness, and the minimum cost was simultaneously employed. Seven decision parameters were length in hot and cold side, frequency of fin, number of fin layers, thickness of fin, fin height, and fin lance length [6]. Gravitational search algorithm was developed from economic point of view. The algorithm was applied to two cases compared to the original data and other algorithms. The total cost could be reduced by 22.3%

as compared to the original data The Gravitational search algorithm could be successfully applied for design optimization [7].

II. Material and Methods

Procedure to design is conducted through some steps. The step is started with input data mass flow rate and temperature both shell and tube side as well as on inlet and outlet respectively. And then calculations are executed to get overall heat transfer coefficient and pressure drops.

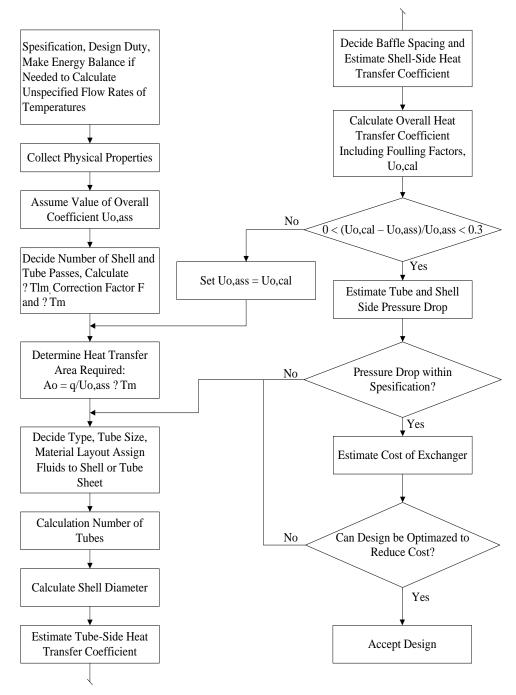


Fig. 2. Design procedure of shell-and-tube heat exchanger [8]

Along calculation processes, assumption and some designer decision are given such as assuming the overall heat transfer coefficient and deciding of some construction type. If the value of overall heat transfer after calculation is less than 30% of the ratio between overall calculated and assumption values of heat transfer while pressure drop does not exceed reasonable limits prescribed, then the design is accepted to be used [8]. Another design may be needed if a designer considers getting a lower cost of heat exchanger.

Procedure to design is conducted through some steps. The step is started with input data mass flow rate and temperature both shell and tube side as well as on inlet and outlet respectively. And then calculations are executed to get overall heat transfer coefficient and pressure drops. Along calculation processes, assumption and some designer decision are given such as assuming the value of overall heat transfer coefficient and deciding of some construction type. If the value of overall heat transfer after calculation is less than 30% of the ratio between overall calculated and assumption values of heat transfer while pressure drop does not exceed reasonable limits prescribed, then the design is accepted to be used [8]. Another design may be needed if a designer considers getting a lower cost of heat exchanger

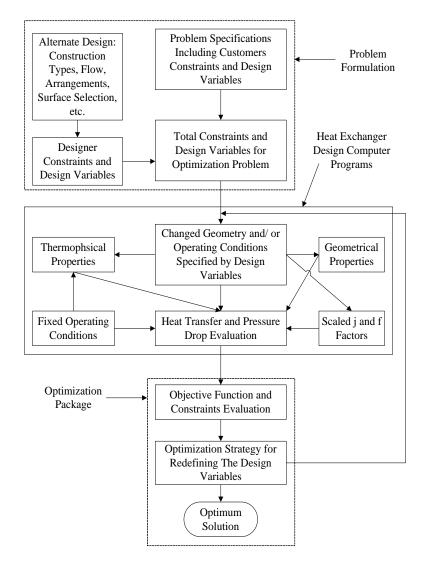


Fig. 3. Methodology for heat exchanger optimization [1]

The program has four variables that are outer tube diameter, tube length, baffle cut to shell inside diameter ratio and baffle spacing to shell inside diameter ratio. Bounds the program for the four variables are described in Table 1. The first bound, Tube outer diameter is taken from BWG standard, which is used correspond to TEMA standard for tube size. The minimum value of outer tube diameter considers cleaning process and vibration of tubes. The cleaning process in the tube can be done with minimum tube size 0.01905 m and vibration also will be reduced using minimum tube size 0.01905 m [9]. The second bound, Range of pipe length depends on space to be expected on size. The third bound, baffle cut to shell inside diameter ratio uses ratio the value ranging from 15% to 45%. It is set to support tubes mechanically against sagging and possible vibration [9]. And the fourth bound, baffle spacing to shell inside diameter ratio uses ratio the value ranging from 20% to 80%. Maximum TEMA standard for baffle spacing is also 80%. It is also used to avoid failure due to tube vibration where it occurs in unsupported tube length more than 80% [9].

Variable	Minimum Value (m)	Maximum Value (m)
Tube outside diameter (do)	0.01905	0.051
Tube length (L)	1	10
Baffle cut to shell inside diameter ratio	0.15	0.45
Baffle spacing to shell inside diameter ratio	0.2	0.8

Table 1. Bounds of variables

The methodology is divided into three main parts which are problem formulation, design and computer program and optimization package, as presented in Fig. 3 Recently, it is possible to get design with minimum cost and satisfied on some constraints by commercial software using optimization methods.

Equations for heat transfer in tube side have many forms. The equation can be selected exactly using a validity statement and Reynolds number. Some correlations heat transfer coefficient in tube side for no phase change process are expressed as follows [10].

For
$$\left(\frac{\text{Re}_{t} \Pr_{t} d_{i}}{L}\right)^{1/3} (\mu / \mu_{w})^{0.14} < 2$$

 $h_{i} = 3.66K_{t} / d$ (1)
For $\left(\frac{\text{Re}_{t} \Pr_{t} d_{i}}{L}\right)^{1/3} (\mu / \mu_{w})^{0.14} > 2$

For $Re_t < 2100$

$$h_{i} = (Kt / di) \ 1.86 \ \left(\frac{Re_{t} \ Pr_{t} \ d_{i}}{L}\right)^{1/3} \ (\mu / \mu_{w})^{0.14}$$
(2)

For $2100 < \text{Re}_t < 10^4$

$$h_{i} = (Kt / di) \ 0.116 \ (Re_{t}^{2/3} - 125) \ Pr_{t}^{1/3} \ (1 + d_{i} / L)^{2/3} \ (\mu / \mu_{w})^{0.14}$$
(3)

For $Re_t > 10^4$

$$h_{i} = (Kt / di) \ 0.027 \ Re_{t}^{0.8} \ Pr_{t}^{0.4} \ (\mu / \mu_{w})^{0.14}$$
(4)

Heat transfer in shell side for no phase change is calculated using Bell-Delaware method. It is more complex, but it is accurate enough. The Bell-Delaware method compares to an ideal tube bank, consider leakage through leakages and bypass flows. So, calculation

of Bell-Delaware method will consider correction factors. Heat transfer in shell side can be found by Eq. 5 [1].

$$\mathbf{h}_{\mathrm{o}} = \mathbf{h}_{\mathrm{id}} \mathbf{J}_{\mathrm{c}} \mathbf{J}_{\mathrm{l}} \mathbf{J}_{\mathrm{b}} \mathbf{J}_{\mathrm{s}} \mathbf{J}_{\mathrm{r}} \tag{5}$$

Total pressure drop is the summation of pressure drop from tube and shell side. Pressure drop in tube side commonly due to frictions and indentations along tubes. Pressure drop for all tubes can be obtained by Eq. 6 [11].

$$\Delta P_{t} = v_{t}^{2} / 2 \left(\frac{4f_{t} L}{d_{i}} + 2.5 \right) n_{p} \rho_{t}$$
(6)

Pressure drop in shell side is calculated using Bell-Delaware method which is evaluated from cross-flow tip baffle to tip baffle. Pressure drop in shell side is commonly due to dividers from baffle and frictions along flow in shell side. Pressure drop in shell side is the sum of pressure drop from the central section, window area and inlet-outlet area considering some correction factors. Pressure drop in shell side can be determined by Eq. 7 [1].

$$\Delta P_{s} = [(N_{b} - 1)\Delta P_{b,id}R_{b} + N_{b}\Delta P_{w,id}]R_{l} + 2\Delta P_{b,id}\left(1 + \frac{N_{r.cw}}{N_{r.cc}}\right)R_{b}R_{s}$$
(7)

The estimated cost of a heat exchanger is got from the summation of investment and operational cost. The total cost can be expressed by Eq. 8 [12].

$$C_{tot} = C_{inv} + C_{op} \tag{8}$$

The investment cost is used as the initial cost to make a shell-and-tube heat exchanger. It can be especially determined for shell material and tube material by Eq. 9, Eq. 10, Eq. 11, Eq. 12 or Eq. 13 [13].

For material (Shell: Carbon Steel and Tube: Carbon Steel)

$$C_{\rm in} = 6411 + 329.7 A^{0.80} \tag{9}$$

For material (Shell: Carbon Steel and Tube: Stainless Steel)

$$C_{\rm in} = 7731 + 372A^{0.85} \tag{10}$$

For material (Shell: Stainless Steel and Tube: Stainless Steel)

$$C_{\rm in} = 8000 + 259.2 A^{0.91} \tag{11}$$

For material (Shell: Carbon Steel and Tube: Titanium)

$$C_{\rm in} = 12821.9 + 562A^{0.92} \tag{12}$$

For material (Shell: Titanium and Tube: Titanium)

$$C_{\rm in} = 16027 + 640 A^{0.93} \tag{13}$$

The operational cost has been used for an operational process for a lifetime of a heat exchanger. Actually, operational cost is used for pumping power due to pressure drop in shell and tube side. Operational cost is calculated considering inflation rate and efficiency of the pump. Operational cost due to inflation rate effects for the lifetime can be determined by Eq. 14 [14].

$$C_{op} = \sum_{k=1}^{n_y} \frac{C_o}{(1+\lambda)^k}$$
(14)

Operational cost for annual current cost is calculated considering operation hours. It can be determined by Eq. 15 [14].

$$C_{o} = P K_{el} \tau$$
(15)

Where pumping power considering efficiency of pump can be calculated using Eq. 16 [14].

$$\mathbf{P} = \left(\frac{\dot{\mathbf{m}}_{t} \,\Delta \mathbf{P}_{t}}{\rho_{t}} + \frac{\dot{\mathbf{m}}_{s} \,\Delta \mathbf{P}_{s}}{\rho_{s}}\right) \frac{1}{\eta} \tag{16}$$

Process of genetic algorithm is started with defined initial parameters. The process continues until maximum number of iteration and satisfy the criteria. A flow process for genetic algorithm is illustrated by flow chart in Fig. 4.

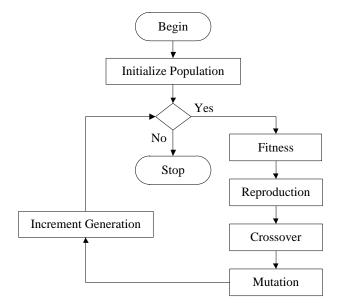


Fig. 4. Principle process of genetic algorithm [15]

Genetic Algorithm is a search algorithm which is built an imitating mechanism of natural selection principles of Darwin for the survival of the fitness. Genetic Algorithm mimics natural process to transmit heredity characteristics from a parent to an offspring by genes in chromosomes. The process will continue from a parent to offspring to get the best individual. The best individual represents an optimal solution [15]. Flow process for Genetic Algorithm is illustrated by flow chart in Fig. 4.

III. Results and Discussions

A. First Case Study

The first case study is a kerosene liquid in shell side and crude oil in tube side. Both shell and tube are made of stainless steel. Energy cost for shell and tube is set as $0.12 \notin kWh$ and interest rate is set as 10% per year. Working hour is set as 7,000 hours/year and the lifetime is set as 10 years with the efficiency of pump 0.7 [11].

Data of fluids and physical properties are known for both stream sides. The data of each stream are mass flow rate, temperature inlet and outlet, density, viscosity, thermal

conductivity, specific heat and fouling resistance. The data of each stream is detailed in Table 2.

	ṁ (kg/s)	Th (°C)	Tc (°C)	ρ (kg/m ³)	μ x 10 ⁵ (Pa.s)	k x 10 ² (W/mK)	Cp (J/kg)	Rf x 10 ⁴ (m ² K/W)
Shell Side: Kerosene	5.52	199.0	93.3	850	40	13	2,470	61
Tube Side: Crude Oil	18.80	37.8	76.7	995	358	13	2,050	61

Table 2. First case study: data of fluids and physical properties [11]

The original design from the first case study uses pattern of square tube arrangement, one shell pass, four tube passes, tube pitch equal to 1.25 of outer tube diameter and baffle spacing equal to 0.24 of inner diameter shell [11]. Optimization of the first case study was carried out by genetic algorithm. Comparison of the result optimization to original data is presented in Table 3.

Parameters	Original Data	Genetic Algorithm
Tube Layout (°)	Square	30
N (Shell)	1	1
Np (Passes)	4	2
Nt (Tubes)	158	202
do (m)	0.025	0.01905
di (m)	0.020	0.01619
Ds (m)	0.539	0.400
Pt (m)	0.031	0.02381
Lbc (m)	0.127	0.133
Lc (m)	-	0.063
L (m)	5.983	5.293
$A(m^2)$	74.21	64.18
$\Delta T lm(K)$	84.55	84.55
F	0.89	0.89
v _t (m/s)	1.523	0.906
v _s (m/s)	0.483	0.615
Gt (kg/m ² s)	1,515.4	901.2
Gs (kg/m ² s)	410.6	522.6
Pr _t	5.6	5.6
Prs	7.6	7.6
Ret	8,468	40,762
Re _s	25,344	24,890
Q (W)	1,441,156	1,441,156
hi (W/m ² K)	1,086	2,112.9
ho (W/m ² K)	978.9	745.1
$U (W/m^2K)$	268.1	309.9
ΔP_t (Pa)	53,195	7,900
ΔP_{s} (Pa)	25,344	16,550
P (W)	1,671	367
C _{in} (€)	21,054	19,438
$\mathrm{C}_{\mathrm{op}}\left(inom{e} ight)$	8,920	1,894
C _{tot} (€)	29,974	21,332

Table 3. Design comparison of the first case study to original data

Optimization process using genetic algorithm has been successfully minimizing total cost on the first case study. Algorithm methods have been decreasing the total cost of the shell-and-tube heat exchanger 28.83% using Genetic Algorithm from the total cost of original data, as mentioned in Table 7. Total cost decreases on the first case study due to decreasing total investment and operation cost. In this case, total operational cost decreases 78.77% using Genetic algorithm. Total investment cost decreases 7.68% using Genetic algorithm from total operational cost and total investment cost of original data.

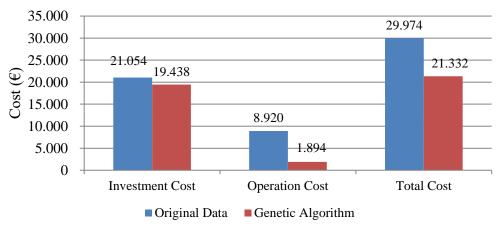




Fig. 5. Cost comparison of first case study

Value of overall, tube side and shell side heat transfer coefficient tends higher than original data. Results of overall heat transfer coefficient increases 15.60% using Genetic algorithm from overall heat transfer of the original data. For heat transfer in tube side, the results increase 94.56% using Genetic algorithm from original data. For heat transfer in shell side, the results increase 23.88% using Genetic algorithm from the original data. Overall heat transfer increases compared to original data because heat transfer area is smaller than the original data. It affects increasing value of heat transfer coefficient in shell and tube side, as presented in Fig. 6.

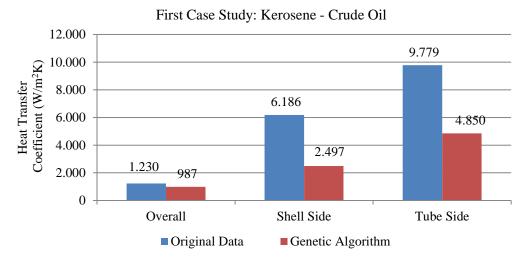
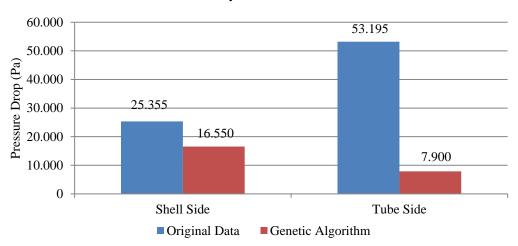


Fig. 6. Heat transfer coefficient comparison of first case study

Pressure drop tends to decrease in the tube side. Pressure drop in tube side decreases 85.15% using Genetic algorithm from the original data. Pressure drops in shell side increases 28.87% using Particle Swarm algorithm from original data. As appears in Fig. 7, It happens because velocity both in tube and shell side is decreased.



First Case Study: Kerosene - Crude Oil

Fig. 7. Pressure drop comparison of first case study

B. Second Case Study

The second case study is a distilled water in shell side and raw water in tube side. Both shell and tube are made of stainless steel. Energy cost for shell and tube is set as $0.12 \notin kWh$ and interest rate is set as 10% per year. Working hour is set as 7,000 hours/year and the lifetime is set as 10 years with the efficiency of pump 0.7 [11].

Data of fluids and physical properties are known from both stream sides. The data of each stream are mass flow rate, temperature inlet and outlet, density, viscosity, thermal conductivity, specific heat and fouling resistance. The data of each stream is detailed in Table 4.

	ṁ (kg/s)	Th (°C)	Tc (°C)	ρ (kg/m ³)	μ x 10 ⁵ (Pa.s)	k x 10 ² (W/mK)	Cp (J/kg)	Rf x 10 ⁴ (m ² K/W)
Shell Side: Distilled Water	22.07	33.9	29.4	995	80	62	4,180	17
Tube Side: Raw Water	35.31	23.9	26.7	999	92	62	4,180	17

Table 4. Second case study: data of fluids and physical properties [11]

Optimization of the second case study was carried out by genetic algorithm. Comparison of the result optimization to original data is presented in Table 5. Optimization process using genetic algorithm has been successfully minimizing total cost on the first case study. Algorithm methods have been decreasing total cost of 52.56% using Genetic Algorithm from the total cost of original data, as mentioned in Table 9. Total cost decreases on the first case study due to decreasing total investment and operation cost. In this case, total operational cost decreases 89.91% using Genetic algorithm. Total investment cost increases 12.39% using Genetic algorithm from total operational cost and total investment cost of original data.

Parameters	Original Data	Genetic Algorithm		
Tube Layout (°)	Triangular	30		
N (Shell)	1	1		
Np (Passes)	2	2		
Nt (Tubes)	160	388		
do (m)	0.019	0.01905		
di (m)	0.0152	0.01619		
Ds (m)	0.387	0.536		
Pt (m)	0.023	0.02381		
Lbc (m)	0.305	0.346		
Lc (m)	5.904	3.030		
L (m)	-	0.156		
$A(m^2)$	56.35	70.22		
$\Delta Tlm(K)$	6.31	6.31		
F	0.94	0.94		
$v_t (m/s)$	2.436	0.887		
v_{s} (m/s)	1.022	0.601		
Gt (kg/m ² s)	2,433.6	885.7		
Gs (kg/m ² s)	1,016.9	597.7		
Prt	6.2	6.2		
Prs	5.4	5.4		
Ret	40,207	15,588		
Re _s	17,155	14,233		
Q (W)	415,137	415,137		
hi (W/m^2K)	9,799	4,849.7		
ho (W/m^2K)	6,186	2,497.2		
$U(W/m^2K)$	1,230	987.1		
ΔP_t (Pa)	65,657	6,052		
ΔP_s (Pa)	88,520	9,849		
P (W)	6,120	618		
C _{in} (€)	18,162	20,413		
C _{op} (€)	31,589	3,188		
$C_{tot}(\epsilon)$	49,751	23,601		

 Table 5. Design comparison of the second case study to original data

Value of overall, tube side and shell side heat transfer coefficient tends lower than original data. Results of overall heat transfer coefficient decreases 19.75% using Genetic algorithm from overall heat transfer of the original data. For heat transfer in tube side, the results decrease 50.51% using Genetic algorithm from original data. For heat transfer in shell side, the results decrease 50.51% using Genetic algorithm from the original data. Overall heat transfer decreases compared to original data because heat transfer area is smaller than the original data. It affects increasing value of heat transfer coefficient in shell and tube side, as presented in Fig. 9.

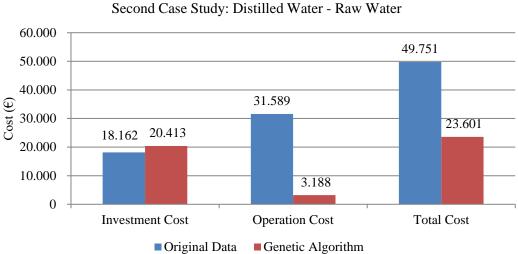


Fig. 8. Cost comparison of second case study

Second Case Study: Distilled Water - Raw Water

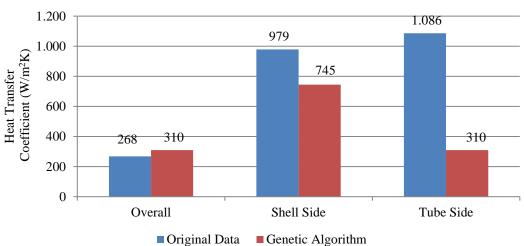


Fig. 9. Heat transfer coefficient comparison of second case study

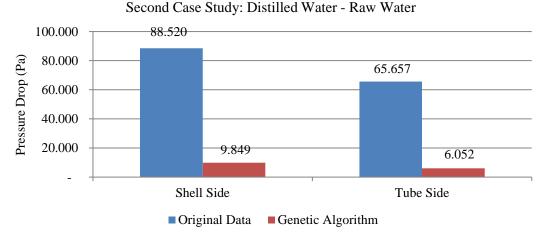


Fig. 10. Pressure drop comparison of second case study

26

Pressure drop tends to decrease in the tube side. Pressure drop in tube side decreases 90.78% using Genetic algorithm from the original data. Pressure drops in shell side decrease 89.67% using Genetic algorithm. As appears in Fig. 10, It happens because velocity both in tube and shell side is decrease.

IV. Conclusion

The program of calculation could be design shell-and-tube heat exchangers tube layout 30°, 45° and 90°. Building codes of efficient algorithm for computational calculation and correspond to TEMA standards has been done, sequences algorithm in computational process were work properly and define TEMA standards into algorithm such as BWG tube standard, minimum value of 1.25 tube pitch to outer tube diameter ratio and maximum value 80% baffle spacing to shell inside diameter ratio. The estimate cost is provided for made of carbon steel, stainless steel, titanium and the combination of their materials. The program has been applied for solving two thermal design shell-and-tube heat exchangers. The first case is a kerosene and crude oil fluids, the results show that program can reduce 28.83% for Genetic algorithm of the total cost from the original data. The second case is a distilled water and raw water, in which the result shows that program can reduce 52.56% for Genetic algorithm of the total cost from the original data.

References

- [1] R. K. Shah, and D. P.Sekulic, *Fundamentals of heat exchanger design*, John Wiley & Sons, Inc., New Jersey, 2003.
- [2] R. V. Rao, A. Saroj, P. Ocloń, and J. Taler, "Design optimization of heat exchangers with advanced optimization techniques: a review", *Archives of Computational Methods in Engineering*, vol. 27, pp. 517–548, 2020.
- [3] M. Dastmalchi, G. A. Sheikhzadeh, and A. Arefmanesh, "Optimization of microfinned tubes in double pipe heat exchangers using particle swarm algorithm", *Applied Thermal Engineering*, vol. 119, pp. 1-9, 2017.
- [4] O. D. Lara-Montañoa, and F. I. Gómez-Castroa, "Optimization of a shell-and-tube heat exchanger using the grey wolf algorithm", in *the 29th European Symposium on Computer Aided Process Engineering*, Elsevier, pp. 572-576, 2019.
- [5] R. V. Rao, and A. Saroj, "Constrained economic optimization of shell-and-tube heat exchangers using elitist-Jaya algorithm", *Energy*, vol. 128, pp. 785-800, 2017.
- [6] H. Zareaa, F. M. Kashkoolic, M. Soltanic, and M. Rezaeianc, "A novel single and multi-objective optimization approach based on Bees Algorithm Hybrid with Particle Swarm Optimization (BAHPSO): Application to thermal-economic design of plate fin heat exchangers", *International Journal of Thermal Sciences*, vol. 129, pp. 552–564, 2018.
- [7] D. K. Mohanty, "Gravitational search algorithm for economic optimization design of a shell and tube heat exchanger", *Applied Thermal Engineering*, vol. 107, pp. 184-193, 2016.
- [8] R. K. Sinnot, and G. Towler, *Chemical engineering design* 6th edition, Elsevier Butterwoth-Heinneman, Oxford, 2019.

- [9] G. F. Hewitt, G. L. S hires, and T. R. Bott, *Process heat transfer*, Begell House, Inc., New York, 2000.
- [10] H. Sadeghzaden, M. A. Ehyaei, and M. A. Rosen, "Techno-economic optimization of a shell and tube heat exchanger by genetic and particle swarm algorithm", *Energy Conversion and Management*, vol. 93, pp. 84-91, 2015.
- [11] J. Yang, S. Oh, and W. Liu, "Optimization of shell-and-tube heat exchangers using a general design approach motivated by constructal theory", *International Journal of Heat and Mass Transfer*, vol. 77, pp. 1144-1154, 2014.
- [12] V. H. Iyer, S. Mahesh, R. Malpani, M. Sapre, A. J. Kulkarni, "Adaptive range genetic algorithm: a hybrid optimization approach and its application in the design and economic optimization of shell-and-tube heat exchanger", *Engineering Applications* of Artificial Intelligence, vol. 85, pp. 444-461, 2019.
- [13] H. Karimi, H.A.D. Ashtiani, and C. Aghanajafi, "Study of mixed materials heat exchanger using optimization techniques". *Journal of Engineering, Design and Technology*, vol. 17, pp. 414-433, 2019.
- [14] T. K. Tharakeshwar, K. N. Seetharamu, and B.D. Prasad, "Multi-objective optimization using bat algorithm for shell and tube heat exchangers", *Applied Thermal Engineering*, vol. 110, pp. 1029-1038, 2017.
- [15] R. K. Arora, Optimization algorithms and applications, CRC Press Taylor & Francis Group, Florida, 2015.