

1 **Economic Performance of Membrane Distillation**
2 **Configurations in Optimal Solar Thermal**
3 **Desalination Systems**

4
5 *Desalination*

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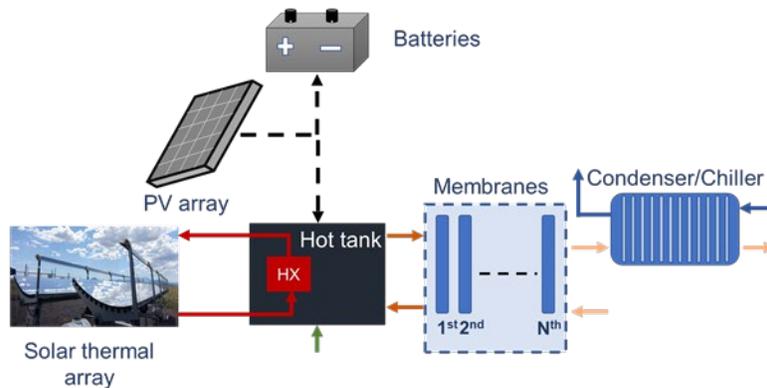
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26 **Abstract**

27 In this study we provide a comprehensive evaluation of the economic performance and
28 viability of solar membrane distillation (MD). To achieve this goal, process models based on mass
29 and energy balances were used to find the minimum cost of water in MD systems. Three MD
30 configurations: direct contact, sweeping gas, and vacuum MD, were compared in terms of
31 economic cost and energy requirements in optimized, solar-driven desalination systems
32 constrained to produce $10 \text{ m}^3 \text{ d}^{-1}$ of distillate from 3.5% or 15% salinity water. Simulation results
33 were used to calculate the water production cost as a function of 13 decision variables, including
34 equipment size and operational variables. Non-linear optimization was performed using the
35 particle swarm algorithm to minimize water production costs and identify optimal values for all
36 decision variables. Results indicate that vacuum MD outperforms alternative MD configurations
37 both economically and energetically, desalting water at a cost of less than \$15 per cubic meter of
38 product water (both initial salt levels). The highest fraction of total cost for all configurations at
39 each salinity level was attributed to the solar thermal collectors—approximately 25% of the total
40 present value cost. Storing energy in any form was economically unfavorable; the optimization
41 scheme selected the smallest battery and hot water tank size allowed. Direct contact MD consumed
42 significantly more energy (primarily thermal) than other MD forms, leading to higher system
43 economic costs as well.

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49 **1. Introduction**

50 Water scarcity is among the most pressing challenges facing humanity, impacting over 2.7
51 billion people worldwide [1-3]. Remote regions lacking public utility infrastructure are among the
52 areas most heavily affected by water scarcity. They are often inhabited by developing communities
53 and marginalized populations, for which the consequences of water shortage are far more reaching.
54 Many such regions experience economic water scarcity, where there is access to non-potable water
55 sources such as saline and hypersaline ground waters, but regional governance lacks the funds to
56 develop these supplies [4, 5].

57 Off-grid desalination is a feasible alternative for mitigating water scarcity in these areas. More
58 specifically, membrane distillation (MD) is a thermal desalination process that can treat high-
59 salinity waters when conventional pressure-driven desalination processes are limited by high
60 pressure requirements [6, 7]. Thermal desalination processes are energy intensive; however, MD
61 can sometimes satisfy energy demands using low grade or waste thermal energy [7-10]. Solar-
62 driven MD is a promising solution for off-grid desalination in the context of the energy-water
63 nexus as it can reduce the overall energy demand by using low grade heat [11, 12].

64 There are four main MD configurations that differ by the mechanism of water vapor collection
65 from the permeate stream [13, 14]: direct contact MD (DCMD), sweeping gas MD (SGMD),
66 vacuum MD (VMD), and air gap MD (AGMD). All use heat to transfer liquid water across a
67 hydrophobic membrane as a vapor. DCMD and AGMD condense the water vapor within the same
68 membrane module; DCMD condenses the vapor into a chilled distillate permeate stream, while
69 AGMD uses a cold surface near the membrane surface. SGMD and VMD, on the other hand,
70 require an external condenser. SGMD uses an air stream to sweep the water vapor to the condenser,
71 whereas in VMD water vapor is transferred from the membrane module to the condenser under
72 vacuum.

73 Each configuration has advantages and disadvantages associated with energy efficiency,
74 water vapor flux, methods of condensation, and thermal energy recovery (Table 1) [15-17]. While
75 DCMD is the simplest configuration, heat transfer through the membrane leads via conduction
76 generally to low thermal efficiency [13]. The introduction of an air gap (AGMD) decreases
77 conductive heat losses and improves thermal efficiency but introduces an additional resistance to
78 mass transfer and reduces water flux [18]. SGMD provides a compromise between DCMD and

79 AGMD; the sweeping gas improves mass transfer over AGMD, while maintaining low conductive
80 heat loss and high thermal efficiency. However, SGMD increases electrical energy requirements
81 for gas transport, and the presence of the sweep gas requires an external condenser to produce
82 liquid water [19]. Additionally, recovering the heat of vaporization of water is more difficult in
83 SGMD and VMD than in DCMD and AGMD. Both AGMD and DCMD can recover thermal
84 energy within the membrane module—AGMD via the cooling plate adjacent to the permeate
85 stream and DCMD via the continuously chilled distillate [20]. Finally, VMD yields high water
86 flux with virtually no heat loss due to conduction [21]. However, like SGMD, VMD requires
87 external condenser and makes thermal energy recovery difficult. Additionally, VMD requires a
88 vacuum pump, increasing the risk of membrane wetting due to lower pressure on the permeate
89 side of the membrane. While the advantages and disadvantages are well known, quantitative
90 comparisons of these configurations are scarce [22, 23], and particularly for off-grid solar MD
91 desalination systems.

92 TABLE 1

93 Several previous studies have modelled off-grid desalination systems to estimate their cost or
94 energy efficiency [24-32]. In these, important factors were sometimes ignored, leading to variation
95 in unit water cost by as much as four orders of magnitude [11, 33]. Most studies did not use
96 comprehensive process models and neglected important considerations such as unsteady operation
97 due to diurnal or seasonal variation in solar irradiance [29]. Some calculated and optimized energy
98 efficiency, while assuming that the most energetically efficient system is also cost-optimal [26,
99 27, 31, 32]. Most importantly, previous studies generally considered only one MD configuration
100 with one energy source and cited cost estimates for other system configurations [24, 27, 31, 32].
101 However, as cost estimates are conditional, a cost comparison among alternate MD systems is only
102 valid if calculated under the same set of circumstances. One study concluded that DCMD was the
103 lowest cost option using solar thermal energy, while AGMD was the most expensive [29]. When
104 the cost of thermal energy was neglected (i.e., based on use of waste thermal energy), VMD was
105 the most cost effective, while AGMD remained the most expensive. Additional work to compare
106 optimized MD systems on the same basis is justified.

107 Herein we compare the economic performances of three membrane distillation configurations
108 — DCMD, SGMD, and VMD — in optimized, solar-driven desalination systems under transient

109 operational conditions. Simulations led to water production costs as a function of key decision
110 variables including the size of individual system components and operational parameters. Non-
111 linear optimization using a particle swarm algorithm identified decision variable values that
112 minimize either water production costs or energy use. Economically and energetically optimal
113 systems were compared to determine the best membrane distillation configuration for point-of-use
114 desalination. Cost distributions of the three MD configurations are exposed through cost and
115 energy analyses on system components.

116 **2. Modeling and optimization methods**

117 **2.1. Solar Membrane Distillation Systems.**

118 The process flow diagrams for the three systems examined (DCMD, SGMD, and VMD) are
119 illustrated in Figs. 1-3. Each system operates in a semi-batch mode by feeding saline water of an
120 initial salinity to a hot water tank (cooling water in). During daily operation, the salinity of the
121 water in the tank increases modestly as the concentrated brine re-enters the tank, where it is mixed
122 with less saline feed water. Mass balance results indicate that over a month of semi-continuous
123 operation, increased salinity in the feed tank does not significantly impact process performance
124 (by lowering the brine-side vapor pressure in the membrane module). Equipment sizes and
125 operational conditions are system decision variables (indicated in italics in the following process
126 descriptions). Stream numbers correspond to those in the figures.

127 In the DCMD system (Fig. 1), the saline feed enters the hot water tank of volume V_{HT} (Stream
128 13). Solar thermal energy is collected using an array of solar thermal collectors of total area A_{th}
129 and transferred to hot water tank via a glycol loop (Stream 1). The heated glycol circulates through
130 a heat exchanger in the hot water tank before returning to the thermal collectors (Stream 2). The
131 heated saline feed passes through an array of N_{mod} flat-sheet membrane modules arranged in
132 parallel at a total flow rate of Q_{hw} (Stream 3). Distilled water is continuously chilled to 25 °C and
133 passed through the membrane modules counter to the direction of the heated saline feed at a flow
134 rate Q_{pw} (Stream 9). Rejected brine is returned to the hot water tank to conserve energy (Stream
135 4), and makeup saline feed water is added continuously to fill up the hot tank (Stream 13).
136 Permeate exits the membrane module in the chilled water stream (Stream 5), where it is cooled by
137 a heat exchanger of area A_{hx} . The cooled distilled (product) water that exits the heat exchanger
138 (Stream 6) is divided into two streams: (i) the distilled water stream used to meet the daily water

139 production requirement (Stream 7) and (ii) the remaining distilled water (Stream 8), which is
140 chilled to 25 °C before being recirculated to the membrane array (Stream 9). Thermal energy is
141 recovered from the hot distilled water by circulating feed (cooling) water (Stream 10) through the
142 heat exchanger at a flow rate of Q_{cw} (Stream 11). The heated cooling water that exits the heat
143 exchanger (Stream 12) is divided into two streams: Stream 13 is used to maintain the volume of
144 water in the hot water tank, and the remaining cooling water (Stream 14) is discarded. Electricity
145 is provided (Line 15) to the three pumps and chiller by a photovoltaic array of area A_{pv} (Lines 17-
146 21). A battery bank sized to operate with a maximum depth of discharge (DOD) provides electrical
147 energy when sunlight is insufficient (Line 16). Note that the chiller was added to maintain ambient
148 temperature at the entrance of the membrane module (distillate side) to support direct comparison
149 to SGMD in which air enters at 25 °C.

150

FIGURE 1

151 In the SGMD system (Fig. 2), saline feed enters the hot water tank of volume V_{HT} (Stream
152 13). Solar thermal energy is collected using an array of solar thermal collectors of area A_{th} and
153 transferred to a glycol loop (Stream 1). The glycol is circulated to a heat exchanger in a hot water
154 tank and returned to the thermal collectors (Stream 2). Heated saline feed is circulated to an array
155 of N_{mod} flat-sheet membrane modules arranged in parallel at a flow rate of Q_{hw} (Stream 3).
156 Permeate diffuses across the membrane into ambient air that passes through the module at a total
157 flow rate Q_{air} (Stream 4). Rejected brine is returned to the hot water tank to conserve thermal
158 energy (Stream 5), and makeup saline feed water continuously replaces permeate to fill up the hot
159 tank (Stream 13). Moist air exits the membrane module (Stream 6) and is sent to a condenser with
160 exchange area A_{cond} . Condensed water and cooled air exit the condenser (Stream 7) and are divided
161 into two streams: (i) the condensed product water (Stream 8), and (ii) the cooled air with any
162 uncondensed water vapor, which is discarded (Stream 9). Thermal energy is recovered from the
163 hot water vapor by circulating cooling (saline feed) water (Stream 10) through the condenser at a
164 flow rate of Q_{cw} (Stream 11). The heated cooling water exiting the condenser (Stream 12) is
165 divided into Stream 13, which is used to maintain the volume of water in the hot water tank, and
166 the remaining cooling water (Stream 14) is discarded. Electricity is provided (Line 15) to the three
167 pumps and blower by a photovoltaic array of area A_{pv} (Lines 17–21). A battery bank with a depth
168 of discharge (DOD) provides electrical power when sunlight is insufficient (Line 16).

169

FIGURE 2

170 In the VMD system (Fig. 3), saline feed enters the hot water tank of volume V_{HT} (Stream 12).
171 Solar thermal energy is collected using an array of solar thermal collectors of area A_{th} and
172 transferred to a glycol loop (Stream 1). The glycol is circulated to a heat exchanger in the hot
173 water tank and returned to the thermal collectors (Stream 2). Heated saline feed is circulated to an
174 array of N_{mod} flat-sheet membrane modules arranged in parallel at a flow rate of Q_{hw} (Stream 3).
175 Permeate is drawn across the membrane under vacuum at a pressure of P_v . Rejected brine is
176 returned to the hot water tank to conserve heat (Stream 4), and makeup saline feed water enters
177 the tank to fill up the hot tank (Stream 12). Water vapor exits the membrane module (Stream 5)
178 and is sent to a condenser of area A_{cond} . Condensed water exits the condenser under vacuum
179 (Stream 6) and is stored in a vacuum tank pending discharge as product (Stream 8). The gas in the
180 head space is removed at a flow rate equal to the volume displaced by the condensed permeate to
181 maintain the vacuum pressure (Stream 7). Thermal energy is recovered from the hot water vapor
182 by circulating the cooling (saline feed) water (Stream 9) through the condenser at a flow rate of
183 Q_{cw} (Stream 10). The heated cooling water exiting the condenser (Stream 11) is divided into
184 Stream 12, which is used to maintain the volume of water in the hot water tank, and the remaining
185 cooling water (Stream 13) which is discarded. Electricity is provided (Line 14) to the three pumps
186 and vacuum by a photovoltaic array of area A_{pv} (Lines 16–20). A battery bank with a depth of
187 discharge DOD again provides electric power when sunlight is insufficient (Line 15).

188

FIGURE 3

189 **2.2. Process Modeling.**

190 A schematic describing the process model and optimization procedure is provided (Fig. 4).
191 The three MD systems were modeled and optimized in MATLAB using methods previously
192 described [7, 33-35]. Here designs were constrained to produce $10 \text{ m}^3 \text{ d}^{-1}$ of purified water.
193 Necessary input parameters included site-specific data (altitude, latitude, longitude, and ambient
194 temperature for the period of operation simulated) and thermodynamic constants (Supplementary
195 material). The process model simulates operation throughout a single day. Necessary input
196 parameters such as solar irradiance and ambient temperature are both time- and site-specific;
197 results are expected to be sensitive to location, regional weather, and raw water characteristics.

198 The optimization and process model procedures consist of the following steps:

199 (i) The optimization algorithm arbitrarily initializes a vector of decision variables and passes
200 them to the process model. The decision variables and limits to the values considered in this study
201 are listed in Table 2. The process model then simulates water production over a 24-h, daylong
202 period of operation. Days are subdivided into three periods of variable length: pre-operation (from
203 midnight to the start of the operating period), operation, and post-operation (from the end of the
204 operating period to midnight). During the pre- and post-operation periods, an energy balance based
205 on environmental heat loss is used to calculate the temperature in the hot water tank as a function
206 of time. Through the pre-operation period the system initiates preheating of the makeup water until
207 the start of the operation period. During the period of operation, solar irradiance, solar-thermal
208 and photovoltaic power, temperature in the hot water tank, battery state of charge, and cumulative
209 water purified are calculated at time intervals of 1 min.

210 (ii) When the 24-hour simulation is complete, the final (midnight) system conditions (e.g., hot
211 water temperature, battery state of charge, and water production) are compared to their initial
212 values. If the conditions are unequal, the initial conditions are reset equal to the calculated final
213 conditions, and the process simulation repeats until the initial and final conditions match, resulting
214 in the simulation of periodic day-to-day operation.

215 (iii) At this point, the process model calculates the total daily permeate production, and the
216 cost calculation algorithm calculates the capital and recurrent or operational costs of the system
217 using the process model outputs. A financial penalty is assigned to solutions that do not meet the
218 production volume constraint. The penalty is of sufficient magnitude to make such solutions sub-
219 optimal [24, 27].

220 (iv) The total present value cost of water purification is calculated and annualized. That is,
221 the equipment cost is calculated from the decision variables and economic heuristics and
222 operational costs from operating conditions before all costs are annualized over a 20-year design
223 period using a 5 percent discount operator. Finally, the breakeven price of water is obtained as the
224 total annualized cost divided by the water produced over a one-year period.

225 (v) The breakeven cost of water is returned to the optimization algorithm. The algorithm
226 adjusts the decision variables, searching for a lower breakeven cost of water, and repeats the
227 process model calculations until that cost can no longer be reduced.

228

FIGURE 4

229

TABLE 2

230 **2.3. Cost Calculation.**

231 Cost functions were developed for equipment components including pumps, blowers,
232 vacuum, chiller, hot water tank, condensers, and solar arrays from manufacturers' data
233 (Supplemental material). Off-the-shelf components such as membrane modules and batteries were
234 given a single unit price (manufacturers' data) so that unit cost was independent of scale.

235 A severe penalty was added to the cost for systems in which the daily water production
236 (calculated from the process simulation) was less than the water production requirement. The
237 application of the penalty is described in detail in our previous studies and it is implemented by
238 applying a coefficient to the cost of water produced (\$1,000) [33, 36]. This coefficient is
239 adequately stringent to ensure that solutions that do not meet the water demand requirement are
240 eliminated by the optimization algorithm. It should be noted that while the production goal and
241 actual production is given in m^3 per day of potable water, the penalty is in US dollars.

242 The objective function represents the amortized cost of water and is set to be minimized
243 during system optimization so that the net present value of the system is equal to zero at the end
244 of the system lifetime. Calculations of membrane and battery replacement, labor, and other
245 operational costs were adapted from Seider et al. [37] and Khayet [11]. The system lifetime was
246 taken as 20 years and the interest rate as 5%. Additional parameters used and assumptions made
247 are detailed in the Supplementary material.

248 **2.4. Process Optimization.**

249 Each MD process was optimized using the particle swarm algorithm in MATLAB following
250 the procedure described in Section 2.2. To investigate the applicability of MD configurations in
251 desalination and treatment of highly saline water, two different feed water salinities were
252 considered: 3.5% and 15%. In all cases, the water production requirement was $10 \text{ m}^3 \text{ d}^{-1}$. The
253 optimization objective was to find the suite of decision variable values that minimizes total present
254 value costs over a 20-year simulation, thus providing the lowest unit cost of water that results in
255 full cost recovery while satisfying the production requirement ($10 \text{ m}^3 \text{ d}^{-1}$).

256 The following optimization procedure was used in all cases:

- 257 1. *Initialization*: initialize values for each decision variable.

- 258 2. *Execution*: run particle swarm process model.
- 259 3. *Adjustment*: adjust initial values of each decision variable to the optimized values found in
260 Step 2.
- 261 4. *Repeat* Steps 2 and 3 until the decision variable values selected by the optimization algorithm
262 equal the input values.

263 The optimization process was provided with (i) an acceptable range of values for each decision
264 variable (Table 2) to serve as solution boundaries and (ii) initial values for each decision variable
265 within that range.

266 The particle swarm algorithm logic mimics the social swarming behavior of birds and insects
267 that follow and converge to single pattern or behavior. The algorithm starts by generating an initial
268 set of individuals, each of which is associated with a location, velocity, and an objective function
269 value within the decision variable space. Each individual is ranked based on the proximity to the
270 best value of a neighboring individual. As the algorithm progresses through iterations, individuals
271 move to new positions that are determined based on the best objective function values of
272 neighboring individuals. Repetition of the algorithm results in the particles swarming around an
273 optimal location. Eventually, the algorithm leads all individuals (now a swarm) to a single,
274 location corresponding to the optimal objective function value. At this point, after the procedure
275 yields the same value three times, the algorithm terminates accepting this value as optimal. This
276 protocol increases the odds of the algorithm reaching a global, rather than local, optimum [38-41].

277 **3. Results and discussion**

278 **3.1. Optimal Membrane Distillation Systems.**

279 A comparative optimization analysis was performed based on the following design
280 constraints: (i) water production: $10 \text{ m}^3 \text{ d}^{-1}$; (ii) location: Leupp, Navajo Nation, AZ, USA; (iii)
281 date: March 20; and (iv) NaCl salinity: 3.5 % and 15 %.

282 March 20 is the spring equinox and represents a median value for both solar irradiance
283 intensity and length of day. Table 3 shows that over a twenty-year lifetime, the optimal VMD
284 system produces water at the lowest cost at $\$14.3 \text{ m}^{-3}$ compared to the optimal DCMD and SGMD
285 systems. The breakeven costs of water produced by the optimal DCMD and SGMD systems under
286 the same conditions were 47% and 24% higher, respectively (Table 3). Surprisingly, the breakeven

287 costs for water produced by VMD and SGMD were insensitive to salt concentration in the range
288 3.5-15 %. However, the same increase in feed water salinity raised the unit cost of water produced
289 by DCMD by about half. At the higher salinity evaluated, the breakeven cost of the DCMD product
290 water was twice that of VMD. VMD and SGMD were also more energetically efficient than
291 DCMD. For the 3.5 % salinity feed solution, the DCMD consumption of energy was more than
292 50 % higher than the unit energy demand of VMD. At 15 % salinity, the difference was more than
293 130 %. In both sets of simulations, the energy consumed by SGMD was on the order of 10 %
294 higher than the VMD energy demand (Table 3). Energy demand in both the VMD and SGMD
295 systems was insensitive to salt concentration of the feed solution.

296

297

TABLE 3

298 The primary source of differences in the economic performance and energy efficiencies of the
299 optimized membrane distillation systems was related to thermal energy consumption. The
300 optimization always selected the smallest hot tank volume permitted— for all configurations and
301 both salinities—because thermal energy storage is not economically advantageous in these
302 applications. As expected, thermal collector areas were much larger in DCMD than in other MD
303 configurations. The optimization algorithm selected the highest brine flow rate allowed in each
304 application (Tables 2 and 3), indicating that the optimum value would be still higher. The
305 importance of maintaining temperature in the brine flow throughout the length of the membrane
306 module is evident. The cooling water flow rate is higher for SGMD and VMD systems as there is
307 a higher requirement for heat dissipation from the permeate stream.

308 For all configurations and salinities, the start and end times of the operational period are
309 approximately the same, with starting time at sunrise and stopping time in late afternoon. This
310 suggests that it is economically and energetically unfavorable to operate on stored thermal or
311 electrical energy. In fact, the battery level of charge was never reduced below 85 %.

312 **3.2. Cost comparison for membrane distillation configurations.**

313 The unit costs of water produced here lie within the range of those previously reported for
314 membrane distillation systems (\$0.40 m⁻³ to \$130 m⁻³) with cost decreasing as water production
315 objectives increase. All studies reporting a cost less than \$1 m⁻³, produced water at a rate >75 m³
316 d⁻¹ [33]. Table 3 and Figures 5-8 support the following observations. Cost contribution of each

317 major piece of equipment and their relevant costs are shown in Fig. 5. The major components are
318 cost to produce thermal and electrical energy, and cost for the membranes and heat exchangers.
319 The cost to produce thermal energy is equal to that of the thermal collector's area plus the cost of
320 the glycol pump. It is interesting to note that membranes do not comprise a significant portion of
321 the costs. While systems vary in terms of average permeate flux and thus require different
322 membrane areas to meet the water production requirement, the membranes do not account for
323 significant cost differences.

324 In all systems, thermal energy expenses constitute the most significant portion of the cost.
325 The thermal component is responsible for the high cost of water of the DCMD system, which is
326 less energy efficient than other forms of MD and requires a larger thermal collector area to satisfy
327 thermal energy requirements.

328 Although the SGMD and VMD systems have similar thermal energy costs, the SGMD system
329 requires a higher condenser area (to condense water from the air stream as opposed to condensing
330 water from a pure water vapor stream in VMD) and more electrical energy to circulate the sweep
331 gas. These two factors are responsible for the modest cost difference between the SGMD and the
332 VMD. Solar thermal collectors make up the largest fraction of cost in all three cases studied (Fig.
333 5) suggesting that process adjustments that increase efficiency in the generation or use of thermal
334 energy would reduce cost. Results of a previous, related study [33], indicated that membranes
335 were the most significant contributor to total cost. Hollow fiber membranes were selected for use
336 in that study, whereas a flat sheet membrane configuration was chosen here. The two
337 configurations differ substantially in performance and particularly price ($\$1,500 \text{ m}^{-2}$ for hollow
338 fiber membranes versus $\$200 \text{ m}^{-2}$ for flat sheet membranes). In this study, the marked reduction in
339 the cost of membranes increased the relative importance of thermal collectors and efficiency in the
340 use of thermal energy. It is possible that with improvements in the efficiency of the current
341 collectors nowadays, the cost of the collectors could be further reduced and thus the membrane
342 component could become an important contributor to the cost again [42]

343 Feed water salinity had little effect on the cost for the SGMD and VMD systems, but a higher
344 impact on DCMD. As discussed earlier, SGMD and VMD are more thermally efficient than
345 DCMD. As a result, in DCMD salinity changes have a higher impact on the local gradient in water

346 vapor pressure across the membrane. That, in turn, impacts the amount of heat required to meet
347 the water production target. SGMD and VMD were not affected by changes in salinity.

348 **FIGURE 5**

349 **3.3 Analysis of Energy Efficiency.**

350 Figure 6 shows the usable thermal energies that enter the hot water tank from the solar thermal
351 collectors (through the glycol loop), the rejected brine, and the preheated makeup saline feed.
352 Thermal energy exits the hot water tank with the heated saline feed (Stream 3, Figs. 1-3) and
353 through environmental losses (not shown). For SGMD and VMD, heat reentering the hot tank with
354 the rejected brines (Stream 5, Fig. 2 and Stream 4, Fig. 3) is significant compared to the demand
355 for solar-thermal energy. In DCMD, however, the energy returned with rejected brine (Stream 4,
356 Figure 1) is small compared to the input of solar-thermal energy (Stream 1, Fig. 1), especially for
357 the 15% salinity case. Each pass of brine through the module transfers much more heat energy in
358 DCMD than in either SGMD or VMD, increasing the capital investment required for thermal
359 energy collection in DCMD. In no case is the thermal energy in makeup water significant
360 compared to other energy flows. Overall, all three optimized systems have small opportunity for
361 heat recovery. However, if forced to recover more water to increase efficiency, the flow rate of
362 makeup water (equal to water production) may become greater than the flow rate of rejected brine.
363 In high recovery systems, recovery of the heat of vaporization may be more important than in the
364 low recovery systems presented here.

365 When the salinity of the feed is 15%, more thermal energy is entering from the solar thermal
366 collectors in the DCMD system (in accordance with our previous findings in Section 3.2 where
367 higher cost is allocated to the area of the solar thermal collectors). For SGMD and VMD, less
368 thermal energy in the hot tank is attributed to the rejected brine than for the 3.5% feed salinity.
369 This observation is also expected as more energy is required to meet the production constraint, and
370 thus, we see reduced energy returned to the hot tank. To better understand this finding, if we were
371 to compare the environmental losses for each configuration at 15%, SGMD and VMD have less
372 thermal energy lost to the environment compared to the DCMD, which translates to SGMD and
373 VMD having better use of the thermal energy compared to DCMD.

374 **FIGURE 6**

375 In Fig. 7 we present the energy exiting the membrane modules for each configuration at the
376 two salinities. Notably, for SGMD and VMD systems, most of the thermal energy exits the module
377 with the rejected brine, which is returned to the hot water tank and conserved; the rest of the
378 thermal energy exits the membrane as water vapor. On the other hand, in the DCMD system, most
379 of the thermal energy is used to heat the permeate carrier (distilled water), and less energy remains
380 in the rejected brine to recycle to the hot water tank. This result explains the low thermal efficiency
381 in the DCMD system: heat is lost to the permeate stream due to conduction through the membrane,
382 whereas heat loss through conduction is negligible in the other configurations. Thus, less thermal
383 energy is returned to the hot water tank in DCMD and more energy from thermal collectors is
384 required.

385 At 15% salinity, DCMD exhibits similar behavior as SGMD and VMD, with most of the
386 thermal energy exiting the module through the rejected brine and a reduction in the amount of
387 thermal energy dissipated by the distilled permeate stream. A possible explanation for this
388 difference with the 3.5% salinity is that the optimization selects a set of decision variables that
389 allows for a higher flowrate of the distillate permeate at the 15% salinity case which in turn
390 translates to lower thermal energy lost to the permeate carrier. In addition, for all three
391 configurations the thermal energy output from the membrane module either with the rejected brine
392 or the permeate is a lower than for the case of the 3.5% salinity. This is a reasonable outcome as
393 more thermal energy is required to meet the constraint of the required amount of water produced.

394

FIGURE 7

395 Figure 8 shows the thermal energy exiting the heat exchangers (for DCMD) or condensers
396 (for SGMD and VMD). In the case of DCMD, thermal energy is dissipated by the permeate stream
397 and is eventually discarded by the chiller before returning to the membrane module (Fig. 1). Due
398 to the high thermal conductance through the membrane, the cooling water, reaches higher
399 temperatures, thus more thermal energy is eventually recycled to the hot water tank (Fig. 6).
400 However, as previously discussed, this is still small in comparison to heat from other sources and
401 does not lead to reduced cost (Fig. 5).

402 In the SGMD and VMD systems, makeup water is preheated by condensing the water vapor
403 exiting the module. In all systems, the thermal energy entering the tank from the preheated makeup
404 water was negligible compared to power inputs from the solar thermal collectors and returning

405 brine. It is concluded that recycling heat with returned brine is more important than recovery of
406 the heat of vaporization in these simulations. Reasons for this are analyzed below. Note that the
407 optimal cooling water flow rate is larger for SGMD and VMD than for DCMD. This is because
408 the heat recovery device in SGMD and VMD have a dual purpose of condensing water vapor and
409 recover any latent heat from the condensation. To maintain an appreciable temperature difference
410 within the condenser and condense more water, a higher cooling water flow rate is selected by the
411 optimization. Nevertheless, this lowers the temperature of the makeup water, and allows for less
412 energy recovery. If lower flowrates would be selected, higher makeup water temperatures could
413 be achieved, but less permeate would be then condensed. This may require additional membrane
414 modules to reach the water production requirement, which in turn would increase the cost. In
415 DCMD, however, the water production rate is not connected to the cooling water flow rate.
416 Therefore, a lower cooling water flow rate was selected, and larger makeup water temperatures
417 were achievable. The optimized DCMD uses more recovered heat than SGMD (Fig. 8) although
418 this does not lead to an overall cost savings.

419 FIGURE 8

420 **4. Conclusions and Outlook**

421 Non-steady-state process models were used to simulated solar driven DCMD, SGMD, and
422 VMD. Following optimization steps, the three forms of membrane distillation were compared
423 based on energy and economic efficiencies. Each process was required to produce $10 \text{ m}^3 \text{ d}^{-1}$ of
424 distillate from 3.5% and then 15% (NaCl) salinity water. Optimization was based on total present
425 value cost or energy consumption using the particle swarm algorithm in MATLAB. In all cases,
426 VMD was more cost effective than SGMD or DCMD. At 3.5% initial salinity, the price of water
427 necessary for full recovery of present value costs (20-year design life, $r = 0.05 \text{ yr}^{-1}$) was \$14.3 per
428 cubic meter. SGMD had a modestly higher electrical energy requirement and higher PV collector
429 area/cost. DCMD proved less energetically efficient, requiring a 50-100 % higher thermal collector
430 area leading to the highest total present value costs. Thermal collectors made up the largest
431 percentage of cost in all cases. Recovery of the heat of vaporization had a much smaller effect on
432 the system energy efficiency than did recycling of rejected brine retained. While this was true for
433 all cases tested, DCMD utilized recovered heat more than SGMD due to the decoupling of the

434 condensation of water vapor and the preheating of the makeup water. However, this did not
435 provide overall cost savings.

436 MD is most advantageous when waste thermal energy is used to treat hypersaline waters.
437 However, the membrane configuration plays an important role on the cost of the system which
438 dictates the applicability of the process for desalination. With our current study we showed that an
439 improvement on membrane configuration can reduce the cost of the process by an order of
440 magnitude ($\$84.7 \text{ m}^{-3}$ from previous study to $\$14.4 \text{ m}^{-3}$ current study). Major equipment costs
441 provide information regarding necessary improvements to make this process economically
442 competitive. Further work is required to anticipate cost savings attributable to the use of state-of-
443 the-art solar-thermal collectors in MD systems. Additional work is also required to examine the
444 effect of the design constraints (here chosen arbitrarily) and the impact of geographic and seasonal
445 variables on the cost of water purified by solar MD.

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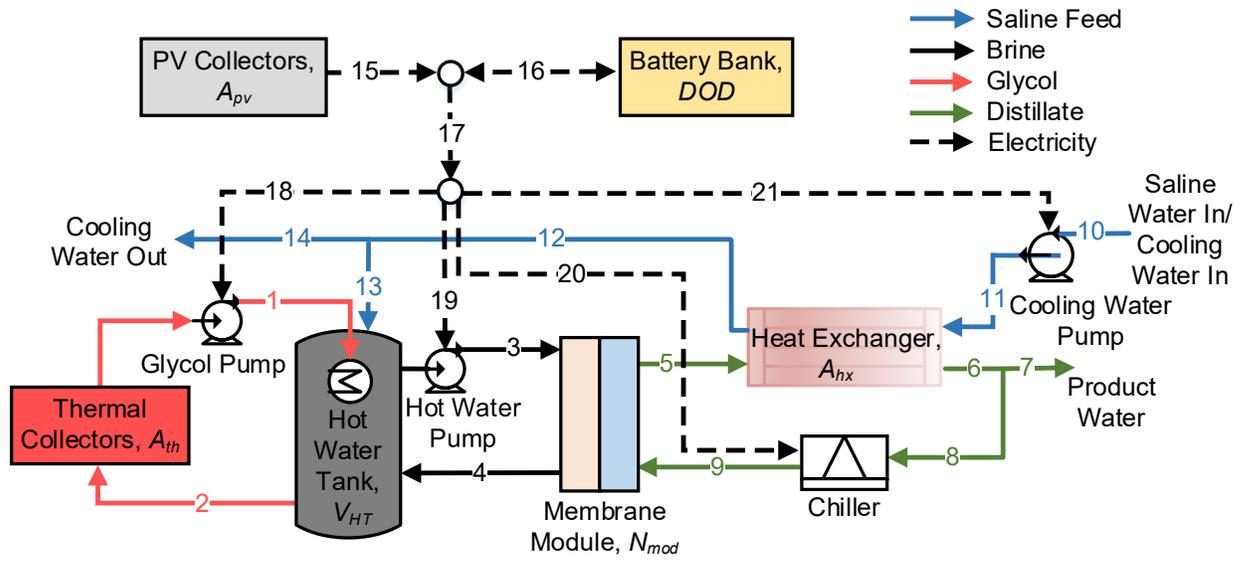


Fig. 1. Flow diagram of solar powered direct contact membrane distillation (DCMD) process. See text.

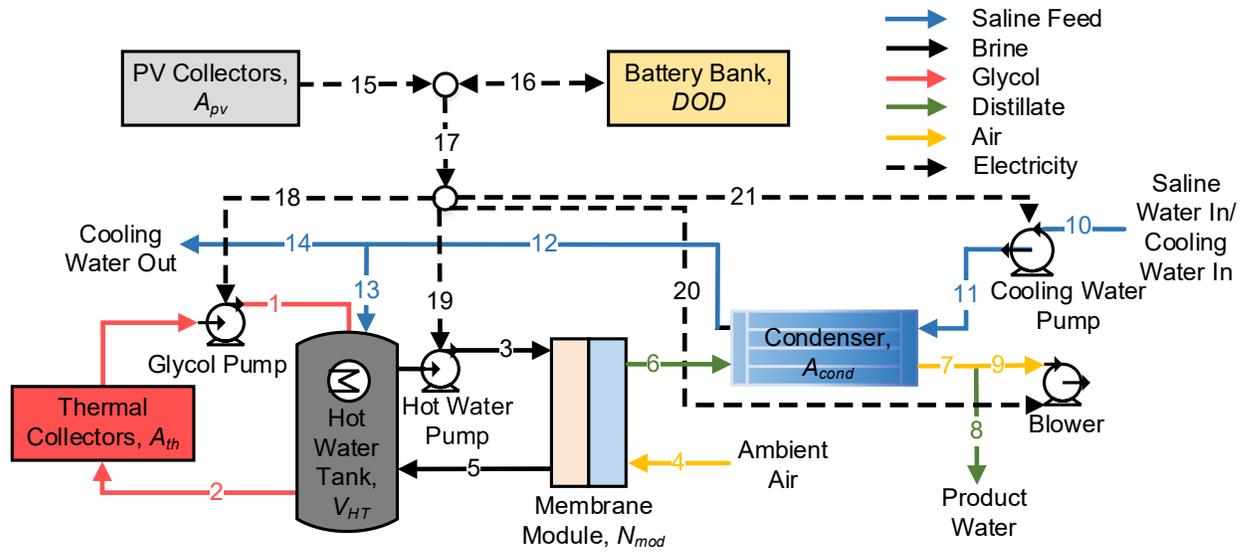


Fig. 2. Flow diagram of solar powered sweeping gas membrane distillation (SGMD) process. See text.

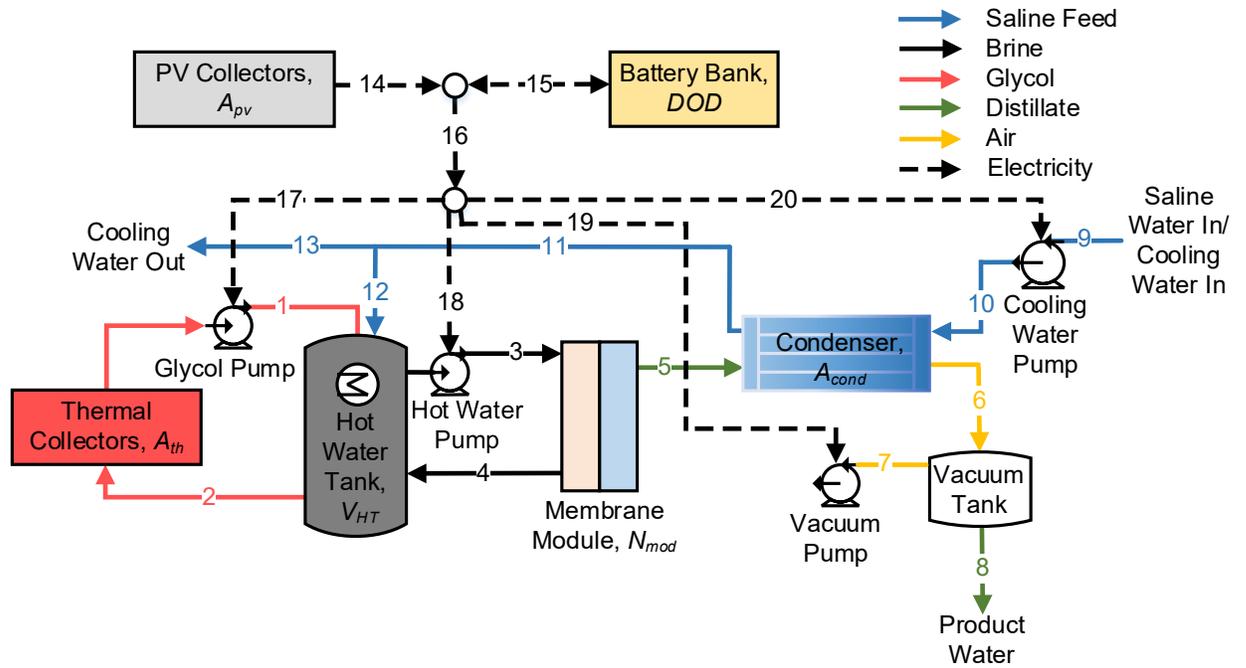


Fig. 3. Flow diagram for solar powered vacuum membrane distillation (VMD) process. See text.

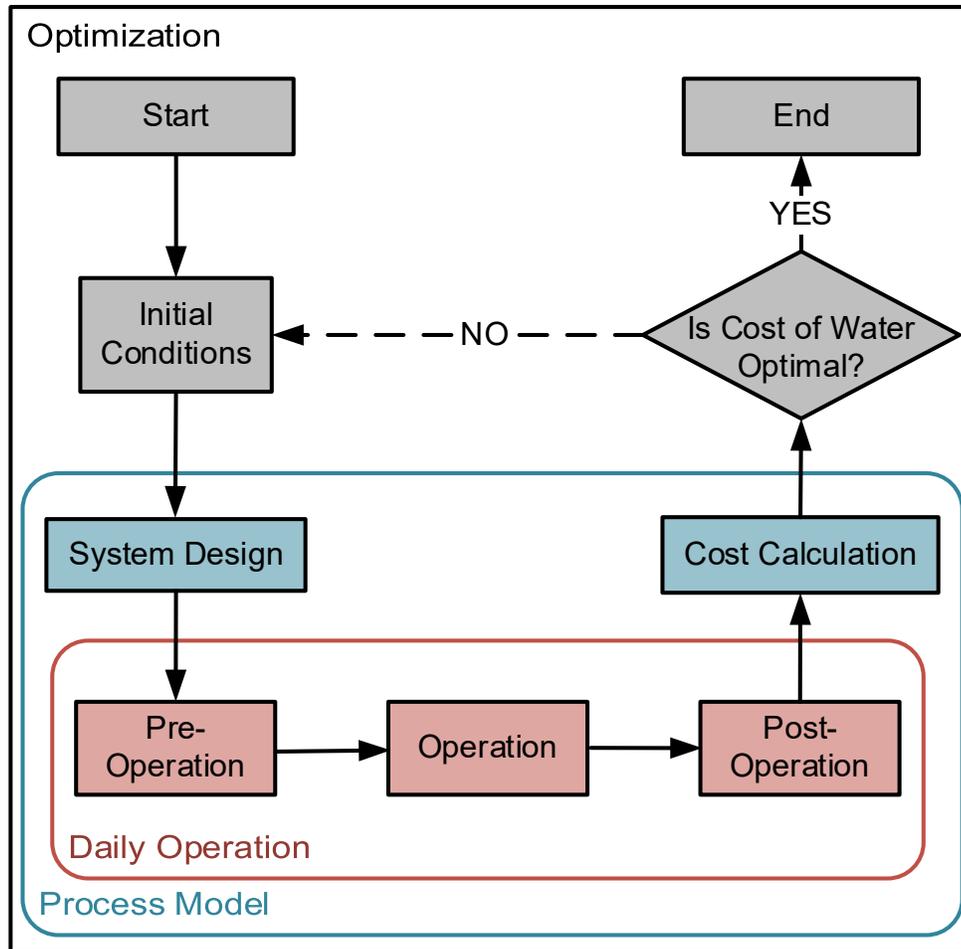


Fig. 4. Overview of process modeling and optimization. The optimization selects decision variable values and passes them to the process model, which calculates the cost of water as a function of the decision variables. The first step of the modeling loop is the decision variable initialization, followed by the process model and the system design, and finally the simulation of daily operation. the daily operation is broken down into three sections: pre-operation (from midnight to the start of the operating period), operation, and post-operation (from the end of the operating period to midnight). Between the water production start and end time, the simulation also calculates the water produced, which varies between configurations. After daily operation is simulated, the final values of the decision variable are compared to their initial values. If they are not equal, the initial conditions are set equal to their final conditions. The process simulation repeats until the initial and final conditions match, resulting in a simulation of day-to-day operation rather than start-up. Finally, the cost of water is calculated. This value is returned to the optimization algorithm. The algorithm repeats multiple times until the values of the objective function swarming around a single optimal location. The algorithm terminates after it produced the same optimal value three times. If the cost is non-optimal, the algorithm selects new decision variable values and repeats the process model until the cost of water is minimized.

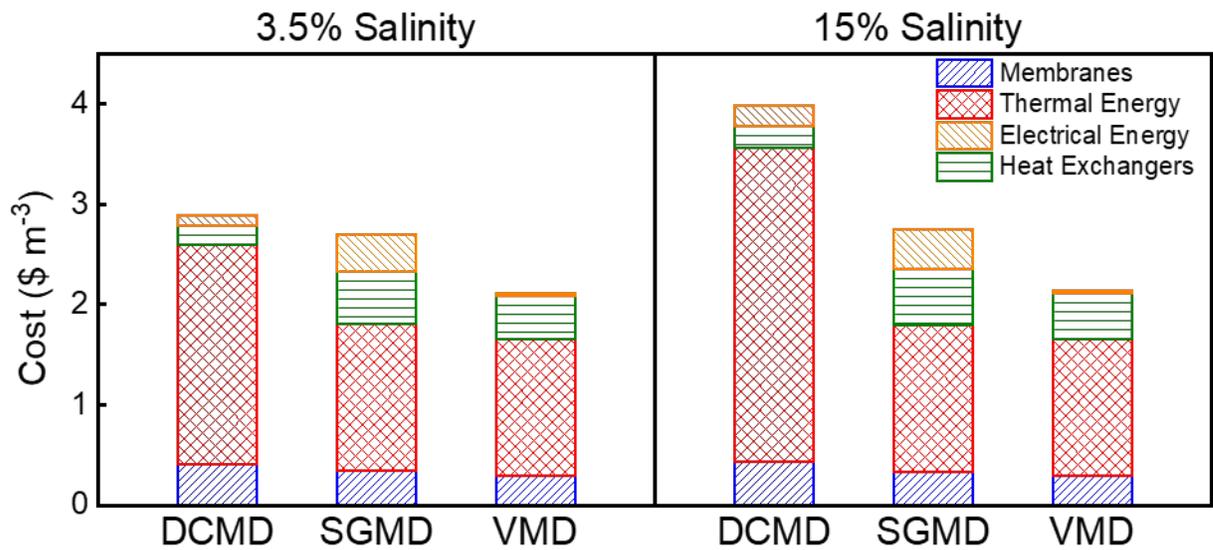


Fig. 5. Selected cost components for optimal DCMD, SGMD, and VMD systems. Note that membrane costs are due to membrane capital and replacement costs; electrical costs include the capital costs of photovoltaic panels and batteries and replacement costs of batteries. Several significant portions of the cost of water are excluded to emphasize the importance of key pieces of equipment. Excluded costs include piping, equipment such as pumps, chillers, and tanks, miscellaneous capital expenses, labor and maintenance expenses, and interest fees.

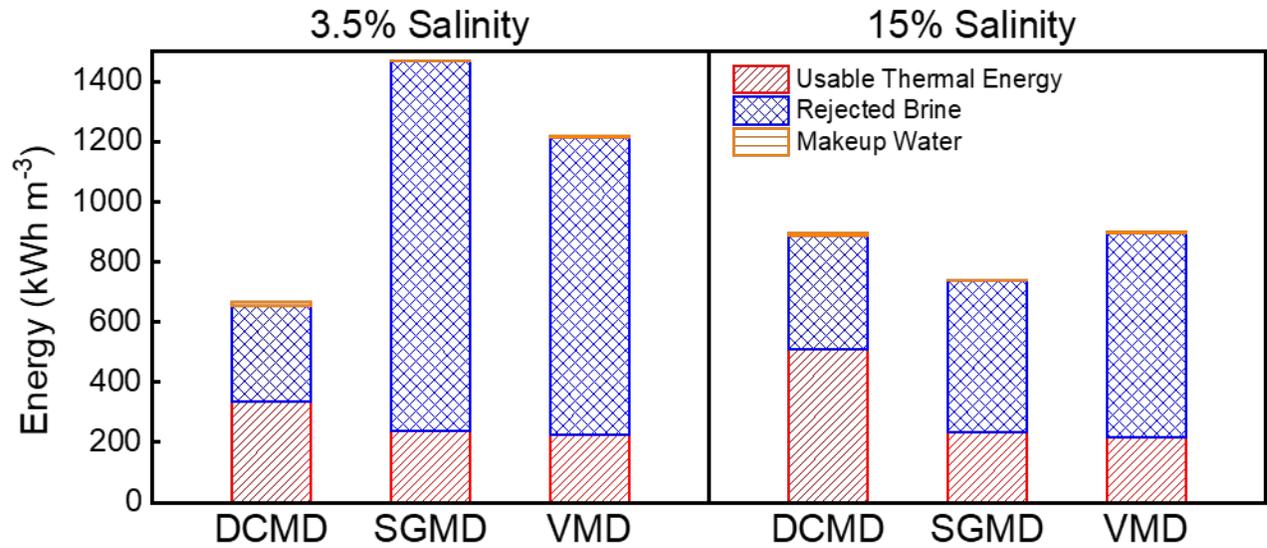


Fig. 6. Thermal energy from various sources entering the hot water tank integrated over time throughout one day of operation in the optimal DCMD, SGMD, and VMD systems. In all three configurations, energy for distillation is provided by the solar-thermal collectors, from rejected brine returned to the tank from the membrane modules, and as makeup water pre-heated using the heat of condensation of permeate (in the heat exchanger in DCMD and the condenser in SGMD and VMD).

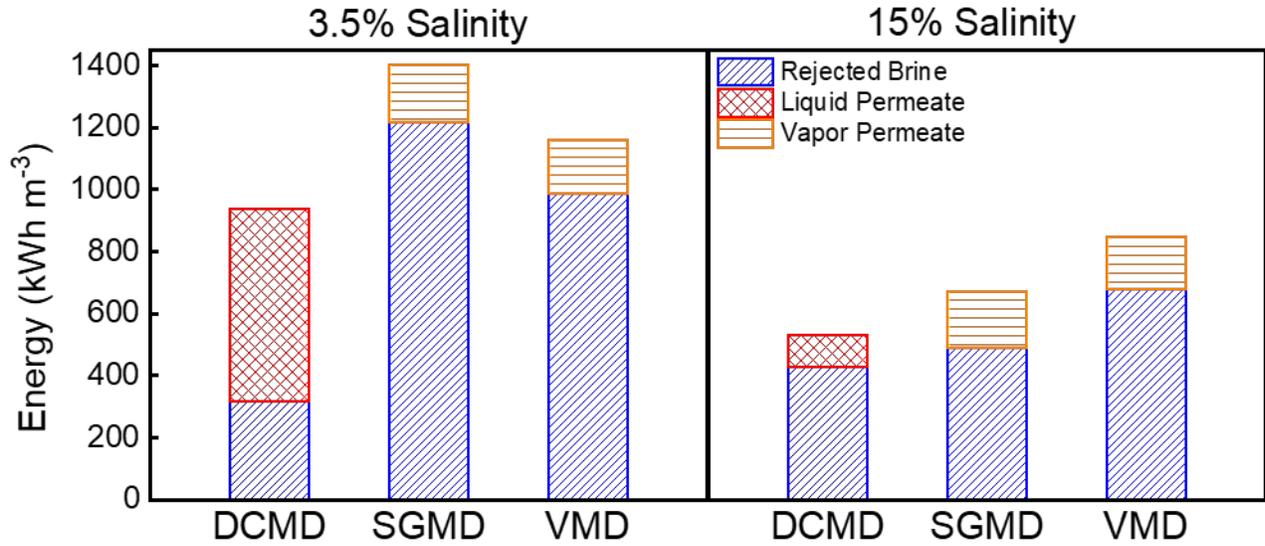


Fig. 7. Thermal energy exiting the membrane modules over one day of operation in the optimal DCMD, SGMD, and VMD systems. In all three configurations, thermal energy is supplied to the modules in the hot brine (Stream 3, Figures 1-3). Thermal energy is lost from the modules in the rejected brine (all systems), the cold contact fluid ((Stream 5, Fig. 1, DCMD) or water vapor (SGMD, Stream 6 and VMD, Stream 5).

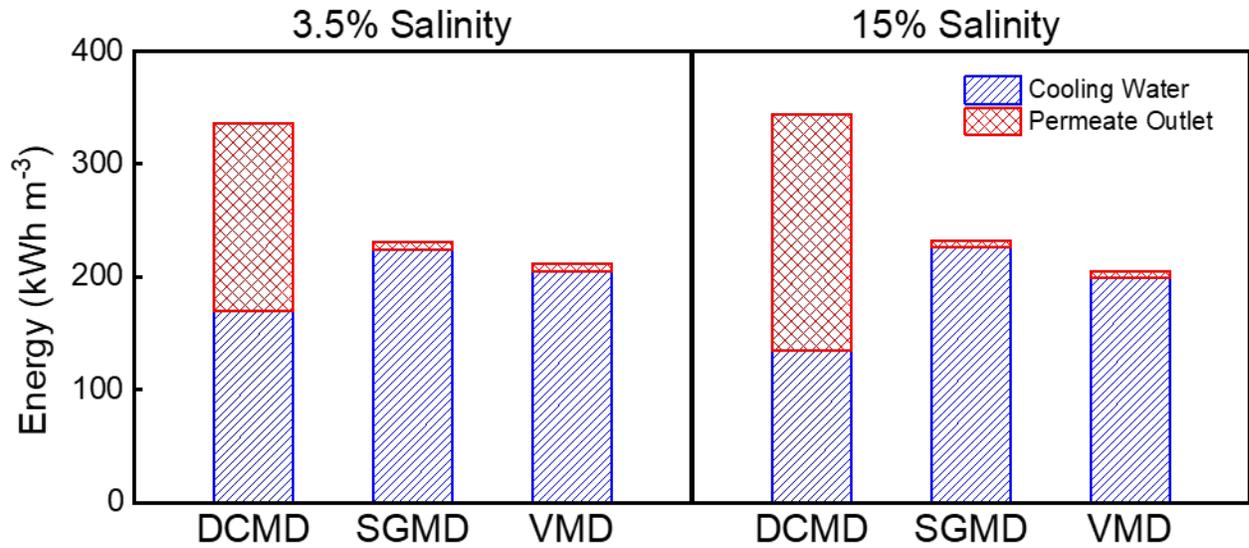


Fig. 8. Thermal energy distribution from the heat exchangers (DCMD) or condensers (SGMD and VMD) integrated over time throughout one day of operation in the optimal DCMD, SGMD, and VMD systems. In the SGMD and VMD systems, nearly all entering energy exits in the cooling water, where a portion of that stream may be fed to the hot water tank to recycle the heat of vaporization of water.

Table 1. Qualitative comparison between membrane distillation configurations across a wide variety of MD performance metrics. [13, 15, 17, 18, 20, 22-25, 32, 43]

Metric	DCMD	SGMD	VMD	AGMD
Conceptual Simplicity	★ ★ ★	★ ☆ ☆	★ ☆ ☆	★ ★ ☆
Thermal Efficiency	★ ☆ ☆	★ ★ ☆	★ ★ ★	★ ☆ ☆
Mass Transfer Efficiency	★ ☆ ☆	★ ★ ★	★ ★ ★	★ ☆ ☆
Heat Recovery	★ ★ ★	★ ☆ ☆	★ ☆ ☆	★ ★ ★
Resistance to Wetting	★ ★ ☆	★ ★ ★	★ ☆ ☆	★ ★ ★

Table 2. Decision variables and the upper and lower bounds of value selection available to the optimization algorithm.

Decision Variables	Upper and Lower Bounds
Membrane Area (m ²)	1 – 100
Hot Water Tank Volume (L)	100 – 2000
PV Collector Area (m ²)	1 – 100
Thermal Collector Area (m ²)	1 – 5000
Brine Flow Rate (L min ⁻¹)	1 – 5
Distillate Flowrate (L min ⁻¹)	0.1 – 50
Air Flow Rate (L min ⁻¹)	0.02 – 200
Vacuum Pressure (kPa)	10 – 90
Cooling Water Flow Rate (L min ⁻¹)	1 – 20
Heat Exchanger Area (m ²)	0.01 – 2
Depth of Discharge	0.01 – 1
Start Time	00:00 (Midnight) – 12:00 (Noon)
End Time	12:00 (Noon) – 00:00 (Midnight)
Water Production Start Time	00:00 (Midnight) – 12:00 (Noon)
Water Production End Time	12:00 (Noon) – 00:00 (Midnight)

Table 3. Decision variable values leading to minimal cost for three different solar MD configurations treating water of two salinities and system performance indicators, including cost of water and specific energy consumption (SEC).

Feed Salinity	3.5 %			15 %		
	DCMD	SGMD	VMD	DCMD	SGMD	VMD
MD Configuration						
Cost of Water (\$ m ⁻³)	21.0	17.7	14.3	29.7	18.6	14.4
Specific Energy Consumption (kWh m ⁻³)	1017	729	650	1536	719	660
Membrane Area (m ²)	39	33	28	42	32	28
Hot Water Tank Volume (L)	100	100	100	100	100	100
PV Collector Area (m ²)	12	50	4	29	51	4
Thermal Collector Area (m ²)	1050	700	650	1500	700	650
Brine Flow Rate (L min ⁻¹)	5	5	5	5	5	5
Distillate Flowrate (L min ⁻¹)	2.5	N/A	N/A	5	N/A	N/A
Air Flow Rate (L min ⁻¹)	N/A	180	N/A	N/A	180	N/A
Vacuum Pressure (kPa)	N/A	N/A	90	N/A	N/A	90
Cooling Water Flow Rate (L min ⁻¹)	2	14	10	2	12	8
Heat Exchanger Area (m ²)	0.05	0.80	0.40	0.05	1.00	0.45
Depth of Discharge	0.90	1.00	0.85	0.9	0.95	0.85
Start Time	5:00	5:00	6:00	5:00	5:20	6:00
End Time	15:30	15:40	15:50	17:30	15:40	16:40
Water Production Start Time	06:00	06:10	06:00	06:00	06:00	06:00
Water Production End Time	15:30	15:40	15:50	16:20	15:30	16:40