Flat Unglazed Transpired Solar Collector: Performance Probability Prediction Approach Using Monte Carlo Simulation Technique

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Abstract: Engineering applications including food processing, wastewater treatment, home heating, commercial heating, and institutional heating successfully use unglazed transpired solar collectors (UTCs). Trapping of solar energy is the prime goal of developing an unglazed transpired solar collector. The UTC is usually developed in and around the walls of the building and absorbs the solar energy to heat the air. One of the key challenges faced by the UTC designer is the prediction of performance and its warranty under uncertain operating conditions of flow variables. Some of the flow features are the velocity distribution, plate temperature, exit temperature and perforation location. The objective of the present study was to establish correlations among these flow features and demonstrate a method of predicting the performance of the UTC. Hence, a correlation matrix was generated from the dataset prepared after solving the airflow over a perforated flat UTC. Further, both strong and weak correlations of flow features were captured through Pearson’s correlation coefficient. A comparison between the outcomes from a linear regression model and that of computational simulation was showcased. The performance probability for the UTC was interlinked with correlation matrix data. The Monte Carlo simulation was used to predict the performance from random values of the flow parameters. The study showed that the difference between the free stream value of temperature and the value of temperature inside the UTC’s chamber varied between 15 and 20 °C. The probability of achieving system efficiency greater than 35% was 55.2%. This has raised the hope of recommending the UTC for drying and heating where the required temperature differential is within 20 °C.

Keywords: solar energy; Monte Carlo simulation; unglazed transpired solar collector (UTC); probability prediction

1. Introduction

Engineering applications including food processing, wastewater treatment, home heating, commercial heating, and institutional heating successfully use unglazed transpired
solar collectors (UTCs). By passing through perforated cladding inside the residential environment, the surrounding air is preheated by the solar collector. Since corrugated UTCs [1–4] are said to be more efficient than the flat type, solar photovoltaic cells are also integrated into the system to provide both power and heat. UTCs are typically used to produce the thermal buoyancy effect, and perforated UTCs are used to create important phenomena such as separation, reattachment, and impingement. When travelling from a perforated section to a flat region, the flow regime changes from laminar to turbulent. Non-homogeneity suction is caused by smaller perforated region features that create a local turbulent flow regime. Some investigations on the perforation spacing, suction flow and velocity field over plate-type UTCs have been carried out by early researchers. Findings from thermodynamic analysis [5], heat transfer analysis and evaluation [6–9], and computational fluid dynamics of flow over UTCs [10] have been reported. To complement the findings on UTCs, both numerical and experimental investigations [11–15] were carried out. Most of these studies were on heat and air flow analysis of photovoltaic thermal systems [16–22]. Thermal modeling and parametric studies to understand the performance of UTCs [23–28] were never easy for early researchers because of challenges posed by the uncertain conditions of incoming airflow [29–31] over a perforated UTC. The performance prediction of the solar energy [32–35] when the system is in use [36–41] is a major challenge under varying conditions of airflow and other key parameters of the UTC.

The performance probability of a flat UTC needs to be studied properly because the UTC is anticipated to function under a great deal of uncertainty when it is put into operation. The constant performance of a UTC in use would be significantly impacted by the substantial fluctuations and dispersion of a few key flow elements.

Investigating the factors for improved performance is a wonderful idea to determine whether a UTC is acceptable. Since the operating conditions of air entering a UTC are usually unknown, an uncertainty analysis was also proposed to understand the performance probability of the UTC. The flow features employed in the computation of UTC performance were compiled into a correlation matrix. For correlation, it is necessary to have a dataset of flow features that have been created by separately solving flows over an unglazed transpired solar collector (UTC), which is intended to gather heat from an outside source.

The present study attempted to establish the correlations between plate temperature, exit temperature, the velocity distribution in the chamber and the location of the UTC perforations. The direct Monte Carlo simulation technique was ideal for achieving the study’s objectives. We sought to predict the performance probability of a UTC in this paper because, to our knowledge, no earlier reports of such attempts have been published. The objective was to compute the likelihood (or probability) of accepting UTC performance when uncertain situations arise during its operation. The performance probability computation was a necessity because, in real-world applications of wind flow and solar energy, many uncertain parameters and their values affect the potential performance of a UTC.

2. Methodology

2.1. Modelling

The cornerstone of the analysis in this study was a set of fundamental governing equations of fluid dynamics—the continuity, momentum and energy equations [42], which are presented as Equations (1)–(4), respectively, below.

Continuity equation:
\[
\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \vec{v}) = 0
\]  

Momentum equation:
\[
\frac{\partial (\rho \vec{v})}{\partial t} + \nabla \cdot (\rho \vec{v} \vec{v}) = -\nabla p + \nabla \cdot (\vec{\tau}) + \rho \vec{g} + \vec{F}
\]  

Energy equation:
\[
\frac{\partial (\rho h)}{\partial t} + \nabla \cdot (\rho \vec{v} h) = \nabla \cdot \vec{q} + \rho \vec{v} \cdot \vec{F}
\]  

Entropy equation:
\[
\frac{\partial S}{\partial t} + \nabla \cdot (\vec{v} S) = 0
\]
where is \( p \) the static pressure, \( \bar{\tau} \) is the stress tensor, \( \rho \vec{g} \) and \( \vec{F} \) are the gravitational body force and external body forces, respectively.

The stress tensor \( \bar{\tau} \) is given below.

\[
\bar{\tau} = \mu \left( \nabla \vec{v} + \nabla \vec{v}^T - \frac{2}{3} \nabla (\vec{v} \cdot I) \right)
\]

where \( \mu \) is the molecular viscosity, \( I \) is the unit tensor, and the second term on the right hand is the effect of volume dilation.

Energy equation:

\[
\rho C_p \left[ \frac{\partial T}{\partial t} + (\vec{v} \cdot \nabla)T \right] = k \nabla^2 T + \varnothing
\]

where \( C_p \) is the specific heat, \( T \) is the temperature, \( k \) is the thermal conductivity of the fluid, and \( \varnothing \) represents dissipation.

For the closure of the unknown Reynolds stress terms (components of total stress tensor) and to model the turbulence effect in the flow, the turbulence models [42] were appropriately chosen for the computational modeling of the flow over the UTC.

The computational domain used in the study is shown in Figure 1a. An area of \( 0.6 \text{ m} \times 0.6 \text{ m} \) with a cavity height of \( 0.15 \text{ m} \) was chosen for the UTC plate. A 1.59 mm diameter perforation was developed for a pitch of 16.89 mm with a wall thickness of 0.86 mm. The upper part of the domain was developed for a height of 0.3 m to accommodate the boundary layer thickness. Details on boundary conditions used for the simulation are mentioned in Table 1. The pattern and orientation of the perforations on the plate (zx-plane) are displayed in Figure 1b. Basically, surface-2 (with perforations) is a wall with heat flux = 600 W/m\(^2\) and acts as an inlet (where suction velocity is defined) to the cavity just below it.

![Figure 1](image-url)
Table 1. Details of Boundary States.

| Surface Number | Boundary Condition Type | Flow Variable                                      |
|----------------|-------------------------|----------------------------------------------------|
| 1              | Inflow                  | Flow enters at 1% turbulence with 1 m/s speed at 298 K |
| 2              | Wall                    | Solar heat flux over wall = 600 W/m²               |
| 3              | Outflow                 | Open to the atmospheric pressure                  |
| 4              | Wall                    | Insulated wall                                     |

A semi-implicit method for pressure-linked equations (SIMPLE) algorithm, an iterative method, was selected in the CFD solver—Ansys® Fluent 20.0, to achieve the solutions of the primitive flow variables. The inflow velocity, \( u \), along surface-1 (indicated as 1 in the domain) varied from 0 to 1 m/s. The direction of this velocity was from left to right of the domain. The suction velocity, \( v = V_s \) (emanating from surface-2) in the cavity, varied from 0.045 to 0.077 m/s, and its direction was vertically downward. All of the boundary conditions are shown in a schematic diagram in Figure 2. The values were decided based on the prevailing experimental conditions and studies in the literature [1,2].

2.2. Mesh Development Analysis

The residuals were calculated under the same boundary conditions with an absolute tolerance value of \( 10^{-6} \) for all variables of flow. A grid independence check was performed
at 0.286 million, 0.2 million, and 0.135 million of grid points for the same computational domain. The mesh analysis was carried out for three different mesh sizes (in each case study), and their values are displayed in Table 2. The plate temperature reported for inlet velocity of 1 m/s (using numerical method) and that noted for the same inlet velocity (using experimental method) differed by 3% only. A similar observation (less than 5% difference) was made for the plate temperature for the inlet velocities of 0.077 and 0.045 m/s for the mesh sizes mentioned below.

Table 2. Mesh Analysis.

| Plate Temperature (K) | Wind Speed and Suction Velocity | Mesh Size | Experimental Data |
|-----------------------|---------------------------------|-----------|-------------------|
|                       | 1 m/s & 0.077 m/s               | Mesh 1    | 314/305.9         | 313.9/306.5       |
|                       |                                 | Mesh 2    | 314.5/305.7       | 314.7/305.0       |
|                       |                                 | Mesh 3    | 314.7/305.0       | 313.9/306.5       |
| Case Study 1          | 1 m/s & 0.0054 m/s              | 315.4/307 | 314.8/307.1       | 313.7/306.9       |
| Case Study 2          | 1 m/s & 0.045 m/s               | 315.4/307 | 314.8/307.1       | 313.7/306.9       |

3. Results and Discussion

Simulation and experimental results were validated for the case studies as mentioned in Table 2. Figure 3a displays the plate temperature, and Figure 3b shows cavity exit air temperature for laminar and turbulent flows using standard k-epsilon (k-\(\varepsilon\)) and shear stress transport (SST) k-omega (k-\(\omega\)) turbulence models. The Eulerian model for CFD simulation was applied in this study.

In the present study, the intent was to understand the mean flow characteristics under turbulent flow conditions, and hence, k-epsilon [42], a two-equation model, was used. To understand the transport of shear stress (indicated in the stress tensor of the momentum equation), the SST k-\(\omega\) model was introduced for closure of the solution. A comparison of numerical results and their validation with experimental data are shown in Figure 3a,b.

The values of temperature obtained using the experimental method and numerical technique compared fairly well. The standard k-epsilon model predicted values of plate temperature in a manner consistent with the experimental values. The predictions of values of temperatures (both plate and cavity exit air temperatures) by laminar flow simulations and turbulent flows (using the above-mentioned turbulence models) were quite complementary in terms of the trends and matched with their corresponding experimental values [1]. The prediction of the y-component of velocity along the flow domain is shown in Figure 3c. The figure shows a scale bar that indicates the maximum and minimum values of mean \(U_y\). It was computed by using the SST k-\(\omega\) turbulence model. The model had better superiority and relevance in predicting the profile of the velocity component. The reason for this decision was that prediction of the velocity component is linked to the transport of shear stress and hence the rate-of-change of deformation of a fluid in an enclosed cavity. The k-\(\varepsilon\) model could also be used to perform the same prediction in this case.
Figure 3. Cont.
with respect to location, a simple regression analysis was performed, and the results were predicted by the laminar model, and the minimum was obtained through the k-ε model. The range of temperature predicted by each CFD model is displayed through a box plot in Figure 4c. To find the best fit line for the differently predicted temperature values, a simple regression analysis was performed, and the results are presented in Figure 4d for the exit temperature and in Figure 4e for the plate temperature. The model fit well and linearly for exit temperature, whereas the non-linearity effect was seen in plate temperature.

Basically, the plots shown in the Figures 4d,e and 5a–e are the results of linear regression between the response and predictor variables. So, the regression line or the best fit line (predicted by the regression method) passed through the actual data points.

Figure 3. Comparison of CFD results and experimental data [1] for flat UTC model: (a) Plate surface temperature [K]; (b) Cavity exit air temperature [K]; (c) CFD Model analysis of Cavity.

Figure 4a shows a comparison of the vertical velocity \( U_y \) profile of air inside the chamber receiving a suction velocity \( V_s = 0.077 \text{ m/s} \). A higher value of \( U_y \) was predicted in the profile with the laminar model as compared to the two turbulent models (k-ε and SST k-ω models). Between \( Y = 0.21 \) and 0.31 (middle of the cavity) the predicted \( U_y \) rose sharply and remained at almost the same value. The non-dimensional quantity \( U_y/U_\infty \) is shown on the horizontal axis where \( U_\infty \) is the freestream velocity of air, and \( Y \), the vertical distance from the base of chamber, is shown on the vertical axis. From \( Y = 0.31 \) to 0.51, the value of \( U_y \) rose. This change was due to the inherent nature of the flow near the walls where friction deterred the growth of velocity (and had no slip condition), and near the surface of the UTC (perforated one), the speed was hugely affected by the suction velocity and hence aided in increasing the magnitude of \( U_y \).

The three CFD models’ predictions of plate temperature are shown in Figure 4b for \( Y = 0.31 \) to 0.5 for the suction velocity of 0.077 m/s. The maximum plate temperature was predicted by the laminar model, and the minimum was obtained through the k-ε model. The range of temperature predicted by each CFD model is displayed through a box plot in Figure 4c. To find the best fit line for the differently predicted temperature values with respect to location, a simple regression analysis was performed, and the results are presented in Figure 4d for the exit temperature and in Figure 4e for the plate temperature. The model fit well and linearly for exit temperature, whereas the non-linearity effect was seen in plate temperature.
Figure 4. Cont.
Figure 4. (a) Comparison of velocity profile $U_y$ of hot air within the chamber predicted by 3 different CFD models for suction velocity, $V_s = 0.077$ m/s. (b) Comparison of plate temperature predicted by 3 different CFD models for suction velocity, $V_s = 0.077$ m/s. (c) Box plots show the range and characteristics of plate temperature predicted by 3 different CFD models for suction velocity, $V_s = 0.077$ m/s. (d) Scatter plot for the distribution of exit temperature $T_{exit}$ (fitted value) versus location. (e) Scatter plot for the distribution of plate temperature (fitted value) versus location.

The exit temperature ($T_{exit}$) and location of perforations (measured from the domain’s inlet) are shown in Figure 4d with a difference; higher temperatures were recorded further from the inlet. The best fit line passed through the actual data points. Residuals between the regression line and the actual distribution of data points decreased over the position of perforation. A similar observation was made for plate temperature in Figure 4e. However, the regression line and actual data distribution trend were unpredictable in this case, and hence, a wider variation in plate temperature ($T_{plate}$) was not visible in the plot. The `lm()` function of the R language (an open-source programming language) was used to carry out regression.

Figure 5a displays the plot between residuals and fitted value over location; for the residual $= 0$, the fitted values were not well behaved and followed a non-linear relationship. Figure 5b displays the plot between residuals and fitted value over exit temperature; the relationship also displayed a non-linear behavior for residual value $= 0$ for velocity in the $y$-direction. Figure 5c is a residual versus fitted plot. Around the residual value $= 0$, the fitted values were very close to the regression line, and the response ($T_e$) from the system
met the regression line (residual is zero); hence, a linear relationship was observed between $T_e$ and the location of the hole.

Figure 5. Cont.
The variation in plate temperature and exit temperature over the y-component velocity is displayed in Figure 5d,e. The regression line determined for the values was slightly linear and the equal error variance was reasonable in this case. The correlation matrix shown in
Table 3 displays the coefficient of correlation, a statistic to measure the linear relationship between any two variables. The variables selected from the study were exit temperature ($T_e$), vertical velocity ($V_y$), experimental exit temperature ($T_{ee}$), numerical plate temperature ($T_p$) and experimental plate temperature ($T_{pe}$). One of the most common ways of measuring a linear correlation between two variables is Pearson’s correlation coefficient \[r\] \[43\]. It is the correlation statistic that measures the statistical strength, which varies from $-1$ to $+1$, between two continuous variables. The formula for $r$ is given below in Equation (5).

$$r_{xy} = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{\sqrt{n \sum x_i^2 - (\sum x_i)^2} \sqrt{n \sum y_i^2 - (\sum y_i)^2}}$$ \hspace{1cm} (5)

The corner values in the matrix are 1s for the same two variables. As an example, a large positive correlation was noted from the matrix when $r = 1.00$ between $T_e$ and $T_{ee}$. Similarly, one small positive correlation was noted, that is, $r = 0.04$ between $T_e$ and $T_p$. A large negative correlation was noted from the matrix when $r = -0.84$ between $V_y$ and location.

The correlation matrix generation was attempted to understand the different variables affecting the performance probability parameters and the overall efficiency of the UTC before it is accepted for any use such as drying. It was necessary to obtain insight into the behavior of the variables concerning changes in another variable that directly or indirectly affected the performance of UTC.

The correlation matrix establishes a statistical foundation \[43\] to understand the linear relationship, and it also provides a clue to narrow down the search for ways to focus on those variables that are sensitive to a small change with respect to the causation of another variable. More importantly, these continuous variables are ingredients of the performance prediction. One of the methods to predict the performance probability of the system (in this case, a UTC) is the Monte Carlo simulation. The above-mentioned variables were integral parts of the simulation \[44,45\].

The Monte Carlo simulation is a mathematical technique (shown in Figure 5f) used to understand the impact of risk and uncertainty in prediction and forecasting models. It is an algorithm that depends upon random numbers (or a random sample of observations or data) to produce the numerical result that is desired from the outcome of the method. Essentially, the concept of the Monte Carlo simulation uses randomness, which can be brought by different distributions of solution space to solve a problem that might be deterministic.

The Monte Carlo technique has been used to predict the overall performance of a flat UTC. The performance of the UTC was studied through its thermal efficiency \[1\], that is

$$\eta = \rho \times V_s \times C_p \times (T_e - T_\infty)/G$$ \hspace{1cm} (6)

where the density of air ($\rho$) is 1.22 kg/m$^3$, suction velocity ($V_s$) is kept between 0.045 m/s and 0.077 m/s, the specific heat capacity of air at constant pressure ($C_p$) is 1.005 kJ/kg·K,
exit temperature ($T_e$) and freestream temperature ($T_{\infty}$) are in Kelvin (K), and solar radiation ($G$) on the UTC is in Watt/m$^2$.

The simulation was run 1000 times to compute the likelihood (41%) of efficiency that could be acceptable for use. The acceptability of use was decided based on the drying capacity of the UTC. The simulation also predicted the likelihood of $T_e$ going greater than 325 K (close to 52 °C) where drying is possible and affects the system’s efficiency [46–48].

The simulation was carried out using the randomness of input variables such as $T_{\infty}$, $T_e$, and $V_s$, and then the efficiency of the UTC was computed for each sample. The efficiency values beyond 35% were accepted for use. So, the variation in system efficiency was between 35% and 65% in this study. The same variation is displayed in the abscissa (G) on the UTC is in Watt/m$^2$.

The drying (the typical range of drying is 10 to 15 °C) chamber varied between 15 and 20 °C. Heated chambers usually can be implemented for drying (the typical range of drying is 10 to 15 °C) purposes. Figure 6d gives the summary of flat UTC performance with an observation that the probability of system efficiency greater than 35% is in Kelvin (K), and solar radiation ($G$) on the UTC is in Watt/m$^2$.

From Figure 6a, it was also noted that the likelihood (or probability) of freestream ($T_{\infty}$) greater than 305 K) influenced about 32.3% of the system efficiency, and a trend line passing inside the variation of data indicated that the average freestream temperature remained at 300 K during the simulation. From Figure 6b, it was also noted that the likelihood (or probability) of suction velocity ($V_s$ beyond 0.08 m/s) influenced about 46.9% of the system efficiency in this study, and a trend line drawn in the collected data points indicated that the suction velocity remained between 0.05 m/s and 0.1 m/s (in Ref. [1]). $V_s$ was between 0.045 and 0.077 m/s.

From Figure 6c, it was also noted that the probability of plate exit temperature ($T_e$ beyond 325 K) influenced about 41.4% of the system efficiency, and a trend line passing through data shown in the plot indicated that the exit temperature remained at 320 K. So, the difference between the freestream value of temperature and that value inside the chamber varied between 15 and 20 °C. Heated chambers usually can be implemented for drying (the typical range of drying is 10 to 15 °C) purposes. Figure 6d gives the summary of flat UTC performance with an observation that the probability of system efficiency greater than 35% is in Kelvin (K), and solar radiation ($G$) on the UTC is in Watt/m$^2$.

Figure 6. (a) Probability of freestream temperature (b) Probability of suction velocity (c) Probability of plate exit temperature (d) Combined Probability graph.
Another observation from the study was that the probability of system efficiency of greater than 35% is 55.2%. This raised the hope of recommending the UTC for drying and heating applications where the required temperature differential is within 20 °C [49–51].

Pair plots are shown in Figure 7 below. From these plots, the relationship among $T_p$, $T_e$, $T_{\text{rise}}$, and the efficiency/maximum efficiency (Eff) ratio in a single plot could be known. It was observed that $T_p$ and $T_e$ were strongly correlated and showed a relationship where both varied linearly. Similarly, $T_e$ and $T_{\text{rise}}$ had a strong linear relationship and the Eff ratio had a linear relationship with $T_{\text{rise}}$.

![Figure 7](image-url) Relationship among $T_p$, $T_e$, $T_{\text{rise}}$, and efficiency/maximum efficiency (Eff) ratio in a flat UTC.

Computational methods were adopted to check the performance of the system with key variables. The fitted value (or regression line) for the response variables (as shown in Figures 4d,e and 5a–e) confirmed the presence of outliers (data points far from the regression line) in the actual data, and hence, these were ignored. The study predicted a strong correlation with the cavity vertical velocity, and a weaker correlation was determined from the Monte Carlo simulation predicted the likelihood of $T_e$ going greater than 325 K (close to 52 °C), and hence, drying with the hot air collected in the cavity (or chamber) could be possible. The simulation also revealed the likelihood of system efficiency being affected by the key decision variables such as suction velocity ($V_s$), plate exit temperature ($T_e$), and freestream temperature ($T_\infty$).

4. Conclusions

From the Monte Carlo simulation, the probability of present system efficiency being influenced by key resolving variables such as freestream temperature ($T_\infty$), suction velocity ($V_s$), and plate exit temperature ($T_e$) was noted. It was also noted that the probability of freestream temperature ($T_\infty$ greater than 305 K) influences about 32.3% of the system efficiency. The probability of suction velocity ($V_s$ beyond 0.08 m/s) influences about 46.9% of system efficiency, while the probability of plate exit temperature ($T_e$ beyond 325 K) influences about 41.4% of system efficiency. The difference between the freestream value of temperature and the value inside the chamber varied between 15 and 20 °C. Another observation from the study was that the probability of system efficiency of greater than 35% was 55.2%. This has raised the hope of recommending UTCs for drying and heating applications where the required temperature differential is within 20 °C. The
predicted average plate exit temperature remained at 320 K, which was also noted during the experiment. The performance probability prediction for the UTC was a key decision-making factor in the design of UTCs—both the flat and corrugated types.

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