Predicting the Extreme Loads in Power Production of Large Wind Turbines Using an Improved PSO Algorithm

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Abstract: Predicting the extreme loads in power production for the preliminary-design of large-scale wind turbine blade is both important and time consuming. In this paper, a simplified method, called Particle Swarm Optimization-Extreme Load Prediction Model (PSO-ELPM), is developed to quickly assess the extreme loads. This method considers the extreme loads solution as an optimal problem. The rotor speed, wind speed, pitch angle, yaw angle, and azimuth angle are selected as design variables. The constraint conditions are obtained by considering the influence of the aeroelastic property and control system of the wind turbine. An improved PSO algorithm is applied. A 1.5 MW and a 2.0 MW wind turbine are chosen to validate the method. The results show that the extreme root load errors between PSO-ELPM and FOCUS are less than 10%, while PSO-ELPM needs much less computational cost than FOCUS. The distribution of flapwise bending moments are close to the results of FOCUS. By analyzing the loads, we find that the extreme flapwise bending moment of the blade root in chord coordinate (CMF_ROOT) is largely reduced because of the control system, with the extreme edgewise bending moment of the blade root in chord coordinate (CME_ROOT) almost unchanged. Furthermore, higher rotor speed and smaller pitch angle will generate larger extreme bending moments at the blade root.

Keywords: extreme load prediction; improved PSO algorithm; bending moment; pitch angle; rotor speed

1. Introduction

With the increased utilization of low wind speed, blade lengths are becoming increasingly longer, so as to capture more energy at low wind speed. By contrast, in order to control costs, the other parts of the wind turbine have largely remained the same. In fact, this treatment implies that the extreme loads, especially the extreme loads at the blade root are constant or increase within the accepted margin of safety during blade design. Therefore, the extreme load at the blade root is one of the most important constraints when designing of low wind speed blades and blade bolts, even though the blade can reliably endure the loads acting on itself. Of course, the extreme load distribution is also critical factor in the evaluation of the safety of blade. However, in this paper we focus on the extreme load prediction at the blade root because it can be easily extended to get the extreme load distribution.

As far as we know, calculating the extreme loads according to the IEC [1] or GL [2] standard is extremely time consuming. Over the last decade, most researchers extrapolated the extreme loads by means of statistical methods which suppose that the distribution of extreme loads at different wind speed meet some statistical models [3–15]. Generally, it is reasonable to estimate the extreme loads...
that occur or a long duration based on the measured or simulated data gathered over a short period. However, these methods are not without their drawbacks. These issues include, which distribution method should be used for the estimation of the short-term extreme load distribution? What are the criteria for the selection of the maximum loads used in the extreme load distribution? And how many time series, simulated or measured, are needed for the proper estimated of the extreme load distribution? Naturally, more data gives better results, however by increasing the number of simulated load cases, the time taken to conduct these simulations is drastically increased. During the design of the blade, the discrete engineering method, has been widely adopted to obtain accurate results. This method gives the extreme loads by simulating every load case according to the IEC or GL standard. Most software programs, such as PHATAS in FOCUS [16] and Bladed [17], all employ this method. This method is more accurate than statistical methods. But the process of this method is also time-consuming and unsuitable for the preliminary design due to excessive load cases required to be simulated during the design life of the wind turbine. For simplicity, many research articles about the design methods regarding blade optimization design select parts of the design load cases (DLCs) to get the extreme load and finally run all the DLCs to verify the results to reduce the overall computational time [18–22]. Even though this process has been used to save time, it takes about 65 h running on a workstation equipped with 40 logical processors [20]. Therefore, this process still represents low efficiency, and it is therefore necessary to propose a new method that can quickly predict the extreme loads for the preliminary design of large-scale wind turbine blades.

In this paper, a simplified method is proposed to quickly predict the extreme loads in power production to facilitate the preliminary design of the blades. The optimal algorithm is used to acquire the extreme loads. Because the blade loads are affected by a number of factors, the solution of predicting extreme loads is a combination of multi-variable, multi-constraint, and multi-object optimal problem. An advanced intelligent optimal algorithm should be found to solve this kind of constraint optimal problem. Compared with the genetic algorithm (GA), the PSO algorithm has its own advantages, such as a simpler and less adjusting parameters. Hence, a simplified extreme load prediction model (PSO-ELPM) by use of an improved PSO algorithm [5] is built and certified by comparing with the results from FOCUS using the discrete method. The results show that the extreme root load errors between PSO-ELPM and FOCUS are less than 10%, while PSO-ELPM needs much less computational cost than FOCUS. The distribution of flapwise bending moments are close to the results of FOCUS. In addition, PSO-ELPM can directly get the status parameters generating extreme loads. This will be useful for control rules design of wind turbines.

The paper is organized according to the following plan. Section 2 gives the details of the PSO-ELPM; Section 3 presents analyses and discusses the corresponding results using the newly proposed model; Section 4 gives the conclusion and some further plans for this model.

2. Model Building

To establish the optimal model for predicting extreme forces, the free variables, constraints, and fitness function need be determined. As we know, the blade loads, especially the flapwise loads, are mainly generated by aerodynamic force. Therefore, the parameters in the equation of aerodynamic force will be mainly analyzed and selected as free variables. Furthermore, the level of gravity is also critical for the edgewise force and will be considered in the solution of fitness function.

In reality, the extreme loads always produce a certain status with few features. According to the IEC and GL standard, the extreme DLCs always selected the rated wind speed \( V_r \), \( V_r + 2 \) and \( V_2 - 2 \) to do the simulation. This is because the extreme loads are always generated around the rated wind speed. However, in this case, the rotor speed always reaches the max value, and the pitch angle usually approximates the minimum value. Therefore, the ability to find the minimum pitch angle and corresponding rotor speed is very important. However, if we calculate the blade load using the dynamic aeroelastic simulation, it will increase the amount of unnecessary calculations because many of the statuses will not generate the extreme value. Therefore, in order to reduce the computational
time, a quasi-static load computation with a consideration of the status parameters of the wind turbine is used in this article. Certainly, storm conditions can be the worst conditions due to large wind speed and inflow angle. Such conditions are not considered in this model, but will be added in the future.

2.1. Design Variables

The corresponding parameters in the aerodynamic force as the main source of the blade load will be selected as the design variables. The aerodynamic force acting on a blade section under the wind speed $V_1$ is shown as Figure 1. The normal force $F_n$ and the tangential force $F_t$ could be expressed as follows:

$$
\begin{align*}
F_n &= \frac{1}{2} \rho V_0^2 c C_n \\
F_t &= \frac{1}{2} \rho V_0^2 c C_t 
\end{align*}
$$

(1)

where $\rho$ is the air density, $c$ is chord length, $C_n, C_t$ are respectively the normal coefficient and tangential coefficient as:

$$
\begin{align*}
C_n &= C_l \cos \phi + C_d \sin \phi \\
C_t &= C_l \sin \phi - C_d \cos \phi
\end{align*}
$$

(2)

where $C_l$ is the lift coefficient, $C_d$ is the drag coefficient, $\phi$ is the inflow angle, from which the attack angle $\alpha$ will be get by subtracting the twist angle $\beta_1$ and pitch angle $\beta_2$, i.e.,

$$
\alpha = \phi - \theta = \phi - (\beta_1 + \beta_2)
$$

(3)

$$
\phi = \arctan\left(\frac{(1 - a)V_1}{(1 + b)\Omega r}\right)
$$

(4)

The inflow velocity $V_0$ in Equation (1) is

$$
V_0 = \sqrt{V_{x0}^2 + V_{y0}^2} = \sqrt{(1 - a)^2 V_1^2 + (1 + b)^2(\Omega r)^2}
$$

(5)

where $a$ is the axial induction factor, $b$ is the tangential induction factor, $\Omega$ is angular speed of the rotor shaft, that is rotor speed in this article.

![Figure 1. Aerodynamic forces acting on a blade section.](image-url)
It can be seen that the load is influenced by lots of variables from Equation (1), such as the wind speed, the chord length, and the pitch angle. Moreover, the wind turbine parameters such as yaw angle $\gamma$, azimuth angle $\psi$ also affect the angle between the wind direction and chord line, so they are all taken into account to obtain the extreme loads. However, some of them are prefixed. Therefore, the rotor speed, wind speed, pitch angle, yaw angle, and azimuth angle will be selected as design variables.

2.2. Design Constraints

After the aerodynamic profile and structure of the blade are fixed, the blade load chiefly depends on rotor speed, pitch angle, and wind speed. However, the above parameters are not independent due to the operation of the control system. Their relationships are nonlinear and changing with the initial condition and time. It is therefore hard to get the specific relationships from them. For the optimal problem, we only need the bound values of these variables and the constraint conditions of the bound values.

2.2.1. Constraints on Pitch Angle

In the following analysis, a 1.5 MW variable speed and variable pitch (VSVP) wind turbine is taken as an example. The constraint conditions will be built by analyzing different load cases, which are simulated with a consideration of the aeroelastic property and the action of the control system by using PHATAS model in FOCUS software.

Constraints between Pitch Angle and Wind Speed

The minimum pitch angle for the wind speed is expressed by $\beta_v$. For VSVP wind turbine, the blade pitch angle will get larger with the increase of the wind speed so as to reduce the rotor speed. Various pitch angles may occur while the wind turbine operates at the same wind speed due to the control system. Hence, obtaining the minimum pitch angle for a wind speed is very difficult, and also the key point of data that is requires so as to confirm the accuracy of the PSO-ELPM.

A variety of load cases have been computed in compliance with GL standard [20], especially the load cases that make it easy to generate the extreme loads, such as the DLC 1.6 and DLC 4.2. Figure 2 gives the different pitch angles for different wind speeds during the simulation of DLC 1.X ($X$ expressed by the number 3, 4, 5, 6, 7, 8, 9) and DLC 4.2, which are computed by FOCUS.

From Figure 2, we can see that the $\beta_v$ varies with wind speeds. When the wind speed is less than 15 m/s, $\beta_v$ can reach the installation angle. But when the wind speed is above 15 m/s, $\beta_v$ becomes larger and larger with the increase of wind speed. Practically, $\beta_v$ will also change with the variation of the aerodynamic profile, blade layers and control strategy at the same wind speed. In this paper, a rough and versatile method is used to get $\beta_v$. Because the extreme DLCs simulate the extreme condition suddenly occurring during the normal operation of wind turbine, the variation of the pitch angle will be not far away from the steady condition by using control system and safety system. Therefore, the $\Delta \beta_2$ is set as the maximum offset value away from the pitch angle under the condition of steady operation. That is to say, if the pitch angle is 10 degrees when the wind turbine normally operates at 15 m/s, the minimum pitch angle is equal to $10 - \Delta \beta_2$ at 15 m/s for all load cases. Practically, $\Delta \beta_2$ may be different for different wind speeds. In order to simplify it, $\Delta \beta_2$ is set as a constant. The $\Delta \beta_2$ can be obtained from the calculation results of all load cases or part of the load cases. For this 1.5 MW wind turbine, $\Delta \beta_2$ is 10.86 degree from part of the load cases computed by FOCUS.

Considering the limitation of the installation angle $\beta_l$, the $\beta_v$ for different wind speeds is shown with red triangles in Figure 2. Based on Figure 2, the $\beta_v$ versus wind speed $V_1$ is divided into three sections.

For the first section, $\beta_v$ is equal to $\beta_l$ when the wind speed is less than $V'$, which is always the rated wind speed. So, it can be expressed as:

$$\beta_m = \beta_l \quad (V_1 \leq V')$$

(6)
For the second section, $\beta_v$ is deduced by the offset $\Delta \beta_2$ from the steady operation value when $V_1$ is between $V'$ and $V_{out}$. For VSVP wind turbine, the steady curve can be expressed as:

$$\beta_v = B_1(V_1)^2 + B_2 V_1 + B_3$$

(7)

where $B_1$, $B_2$, and $B_3$ are constant and can be obtained by quadratic polynomial fitting. Data for fitting, which are assessed by simulating the steady operation of the wind turbine using FOCUS, are expressed as green circle in Figure 2.

Therefore, $\beta_v$ can be expressed as:

$$\beta_v = B_1(V_1)^2 + B_2 V_1 + B_3 - \Delta \beta_2$$  \hspace{1cm} (V' < V_1 \leq V_o) \hspace{1cm} (8)

Besides, $\beta_v$ cannot be less than $\beta_l$, so

$$\beta_v = \beta_l \hspace{1cm} (\beta_v < \beta_l)$$ \hspace{1cm} (9)

For the third section, $V_1$ is larger than $V_o$. Under this condition, the wind turbine exceeds the normal operational range. The safety system will work and make the pitch angle change to feather. It will not stop until the pitch angle reaches the allowed maximum value, which is around 90 degrees. So the relationship of $\beta_v$ versus $V_1$ is assumed as a straight line, which is through the point $(V_o, \beta_o)$ with the slope of pitch rate $K_p$, from $V_o$ to the maximum wind speed. Then, the $\beta_v$ can be expressed as:

$$\beta_v = \beta_o + K_p(V_1 - V_o)$$ \hspace{1cm} (10)

The point $(V_o, \beta_o)$ also matches Equation (8) to guarantee the continuity of the value of $\beta_v$. That is,

$$\beta_o = B_1(V_o)^2 + B_2 V_o + B_3 - \Delta \beta_2$$ \hspace{1cm} (11)

Constraints between Pitch Angle and Rotor Speed

The minimum pitch angle for rotor speed is expressed by $\beta_r$. Figure 3 depicts the different pitch angles for different rotor speeds during the simulation of DLC 1.X ($X$ expressed by the number 3, 4,
5, 6, 7, 8, 9) and DLC 4.2. EOG1_Rated means that the data are obtained from the simulation of an extreme load case, which of the inflow condition is an extreme operational gust in one year (EOG1) at rated wind speed. From Figure 3, it is clear that the $\beta_r$ has different values for different rotor speeds. When the rotor speed is less than 1.82 rad/s, $\beta_r$ can reach the installation angle. But when the rotor speed is above 1.82 rad/s, $\beta_r$ increases with the rotor rise of speed.

However, it is very complicated to accurately attain $\beta_r$. Based on Figure 3, the $\beta_r$ versus wind speed $V_1$ is divided into two sections. For the first section, $\beta_r$ equal to $\beta_l$ when the rotor speed is less than $\Omega'$, which is always the rated rotor speed $\Omega_r$. It can be expressed as:

$$\beta_r = \beta_l (\Omega < \Omega')$$  \hspace{1cm} (12)

For the second section, $\beta_r$ approaches a straight line when the rotor speed is above $\Omega'$ and it can be expressed as

$$\beta_r = C_1 \Omega + C_2 \hspace{1cm} (\Omega \geq \Omega')$$  \hspace{1cm} (13)

where $C_1$ and $C_2$ are the constant and can be obtained from a linear fitting, based on the data shown in the red triangle in Figure 3. These data are obtained by simulating load case with inflow conditions satisfied by EOG1 at rated wind speed, using FOCUS. It also can be experientially given.

Certainly, $\beta_r$ cannot be less than $\beta_l$, so

$$\beta_r = \beta_l \hspace{1cm} (\beta_r < \beta_l)$$  \hspace{1cm} (14)

**Figure 3.** Pitch angle versus rotor speed in some load cases.

**Brief Summary**

According to the above analysis, we can get the minimum pitch angle for a given rotor speed and wind speed, named as $\beta_m$, can be defined as

$$\beta_m = \max(\beta_r, \beta_v) (V_1 \leq V_o)$$  \hspace{1cm} (15)

$$\beta_m = \beta_v \hspace{1cm} (V_1 > V_o)$$  \hspace{1cm} (16)
Therefore, for all the rotor speeds and wind speeds, the corresponding pitch angle should not be lower than $\beta_m$, that is
\[ \beta_2 \geq \beta_m \quad (17) \]

In addition, the pitch angle should be below the allowed maximum value $\beta_u$, which is around 90 degrees. Therefore, the pitch angle should satisfy the following inequality constraint.
\[ \beta_u \geq \beta_2 \geq \beta_m \quad (18) \]

### 2.2.2. Other Constraints

For VSVP wind turbine, there is always a maximum rotor speed $\Omega_b$, where the safety system will work to limit the rotor speed. Certainly, because of the delay of the actuator, the rotor speed still has a certain incremental value $\Delta \Omega$. However, it is difficult to fix. According to the known results from the simulation of FOCUS, it is no more than 0.2 times the rated rotor speed. Therefore, the rotor speed may get the maximum value as
\[ \Omega_u = \Omega_b + \Delta \Omega = \Omega_b + 0.2 \Omega_r \quad (19) \]

In this article, we used the maximum rotor speed appeared in FOCUS analysis. In Section 3, we will discuss the effects of parameter $\Delta \Omega$ on the extreme load. Moreover, the range of the other variables should be confined in compliance with the wind farm class and the design requirements of the wind turbine.

\[
\begin{align*}
\Omega_l & \leq \Omega \leq \Omega_u \\
V_l & \leq V \leq V_u \\
\gamma_l & \leq \gamma \leq \gamma_u \\
\psi_l & \leq \psi \leq \psi_u
\end{align*}
\quad (20)
\]

### 2.2.3. Summary of Constraints

According to the above analysis, Equations (18) and (20) are finally used to limit the change of the design variables. In order to get the $\beta_m$ in Equation (18), Equations (6), (8)–(16) are needed, we can see that there are many parameters that need to be fixed in these equations. Fortunately, some of the parameters in the above equations are given before can begin the blade design. They are $\beta_u$, $\beta_l$, $K_p$, $\Omega_l$, $\Omega_b$, $V_l$, $V_u$, $\gamma_l$, $\gamma_u$, $\psi_l$, $\psi_u$. How to get the unknown parameters except $\Delta \Omega$ is given below in detail.

**Step 1:** We conduct the steady simulation to get the rotor speed, pitch angle, and output power changing with time.

**Step 2:** According to the steady simulation, some parameters can be obtained, such as $V_r$, $\Omega_r$, and $V_o$. Following this, $V'$ is obtained from power versus wind speed (see (a) in Figure 4).

**Step 3:** Based on the pitch angle changing with wind speed, we can get the coefficients in Equation (7) by quadratic polynomial fitting (see (b) in Figure 4).

**Step 4:** By computing the extreme load case with the inflow wind condition, which is satisfied EOG1 at rated wind speed, using FOCUS, we can obtain the data in Figures 2 and 3. Then $\Delta \beta_2$ and $\Omega'$ can be obtained using the following method (see Figure 5).

**Step 5:** We can get the parameter $\beta_o$ according to Equation (11).

**Step 6:** From the simulation results in step 4 (see the red triangles in Figure 3), we can access the coefficients in Equation (13) by linear fitting.

### 2.3. Design Objects

For the large-scale blade used in the low wind speed area, the extreme load, especially the extreme bending moment at the blade root plays a key role in the blade design. It not only influences the blade reliability, but also involves the security of other equipment, such as the pitch motor and connecting
bolts. Hence, for the single objective optimization, the extreme moment in one direction at a given section is selected as the design object. For the multi-objective optimization, the design objects can be defined as the different direction bending moments at the same section, or the same direction bending moment at different sections, or the different direction bending moments at different sections.

Figure 4. (a) Power and (b) pitch angle changing with wind speed in steady simulation.

To compute the bending moments at different sections, the blade element moment (BEM) theory is chosen to obtain the aerodynamic force $F_a$, which are $F_n$ and $F_t$ in Equation (1). From Equations (1)–(5), we can see that the induction factors of $a$ and $b$ are the only unknown numbers, while the other parameters are known or the design variables. In this article, the traditional iteration algorithm for BEM is used to obtain them with considering the tip loss correction and root loss correction. It is as follows.

Step 1: Supposing initial values for $a$ and $b$, such as $a = 0, b = 0$;
Step 2: Using Equation (4) to get the inflow angle;
Step 3: Using Equation (3) to get the attack angle;
Step 4: Computing the blade loss factor $F = F_{tip}F_{hub}$;
Where $F_{tip}$ is tip loss factor, $F_{hub}$ is root loss factor. They are shown as

$$F_{tip} = \frac{2}{\pi} \cos^{-1} e^{-\frac{R(r-R^{-1})}{2r}}$$  \hspace{1cm} (21)
\[ F_{\text{hub}} = \frac{2}{\pi} \cos^{-1} e^{-\left(\frac{B(r-R_{\text{hub}})}{2\pi R_{\text{hub}} \sin \phi}\right)} \]  

(22)

where \( B \) is the number of the blade, \( R \) is the rotating radius of rotor, \( R_{\text{hub}} \) is the rotating radius of hub, \( r \) is the rotating radius of blade element.

Step 5: Computing the thrust coefficient of the rotor \( C_T \):

\[ C_T = \frac{\sigma (1-a)^2 C_n}{\sin^2 \phi} \]  

(23)

where \( \sigma \) is rotor solidity and it is shown as

\[ \sigma = \frac{Bc}{2\pi r} \]  

(24)

Step 6: If \( C_T \geq 0.96F \), the new induction factor \( a' \) will be computed by Equation (25):

\[ a' = 1 - k_1 + \frac{k_1^2 - 4k_2(k_0 - C_T)}{2k_2} \]  

(25)

where \( k_2 = 1/0.18 - 4F \), \( k_1 = -(0.8/0.18 - 4F) \), \( k_0 = 0.16/0.18 \) or else, the new induction factor \( a' \) will be computed by Equation (26).

\[ a' = \left[ 1 + \frac{4F \sin^2 \phi}{\sigma C_n} \right]^{-1} \]  

(26)

Step 7: The new induction factor \( b' \) will be computed by Equation (27):

\[ b' = \left[ \frac{4F \sin \phi \cos \phi}{\sigma C_t} - 1 \right]^{-1} \]  

(27)

Step 8: Computing the error \( \Delta a = |a - a'| \) and \( \Delta b = |b - b'| \);

Step 9: If \( \Delta a \leq 0.001 \) and \( \Delta b \leq 0.001 \), end the iteration and use \( a', b' \) and Equations (1)-(5) to get the \( F_n \) and \( F_t \), that is \( F_n \) or else, return to step 2 use new induction factor \( a' \) and \( b' \).

Moreover, the gravity \( F_g \) and centrifugal force \( F_c \) at different section is also taken into account. They are shown as follows

\[ F_c = m_k \Omega^2 r \]  

(28)

\[ F_g = m_k g \]  

(29)

where \( m_k \) is section mass per length, \( g \) is gravity acceleration.

Then, we can obtain the force \( F_k \) at a certain section by coordinate transformation.

\[ F_k = F_c + F_g + F_a \]  

(30)

Supposing that \( F_k \) is not changing along one blade element, the forces and moments at a section in a certain coordinate can be solved by coordinate transformation and integration.

Finally, the flow chart of the blade load calculation model is given below (see Figure 6).

2.4. Building the PSO-ELPM Model

In this article, an improved PSO algorithm [5] developed by the authors is employed to evaluate the extreme loads. This optimal algorithm has a better optimization capability than the original PSO algorithm. It is suitable for multi-variables, discontinuous, and multi-objects problem. Finally, by combining the blade load calculation model and the improved PSO algorithm, a simplified extreme load predicting model (PSO-ELPM) is built. The flow chart of PSO-ELPM is given below (see Figure 7).
Step 8: Computing the error $\Delta a$ and $\Delta b$;
Step 9: If $0.001 \leq \Delta a$ and $0.001 \leq \Delta b$, end the iteration and use $a'$ and $b'$ and Equations (1)–(5) to get the $nF'$ and $tF'$, that is $aF'$ or else, return to step 2 use new induction factor $a'$ and $b'$.

Moreover, the gravity $gF'$ and centrifugal force $cF'$ at different section is also taken into account. They are shown as follows

\[ \omega \frac{d^2 kF_m}{c} r \left(28\right) \]
\[ g \frac{d^2 kF_m}{c} \left(29\right) \]

where $\omega km$ is section mass per length, $g$ is gravity acceleration.

Then, we can obtain the force $kF'$ at a certain section by coordinate transformation.

\[ kC_g aF' + F_m \left(30\right) \]

Supposing that $kF'$ is not changing along one blade element, the forces and moments at a section in a certain coordinate can be solved by coordinate transformation and integration.

Finally, the flow chart of the blade load calculation model is given below (see Figure 6).

### 2.4. Building the PSO-ELPM Model

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Figure 6. Flow chart of the blade load calculation model.

Figure 7. Flow chart of the particle swarm optimization-extreme load prediction model (PSO-ELPM).
3. Results and Discussion

The model proposed in the paper was applied to a 1.5 MW VSVP wind turbine and a 2.0 MW VSVP wind turbine to predict the bending moments at the blade root. The corresponding results were compared with those computed by FOCUS [12] to validate the model. The particular constraint conditions used in the PSO-ELPM is introduced firstly.

(1) For the 1.5 MW VSVP wind turbine, the constraint conditions were

\[
\begin{align*}
\beta_m &= \max(\beta_r, \beta_v) \quad (V_1 \leq V_o) \\
\beta_m &= \beta_v \quad (V_1 > V_o)
\end{align*}
\]  
(31)

where \( \beta_v \) satisfies the following expressions.

\[
\begin{align*}
\beta_v &= -1 \quad (V_1 \leq 11 \text{ m/s}) \\
\beta_v &= 22.674 - 10.86 + 5(V_1 - 25) \quad (V_1 > 25 \text{ m/s})
\end{align*}
\]  
(32)
(33)

and \( \beta_r \) satisfies the following expressions.

\[
\begin{align*}
\beta_r &= -1 \quad (\Omega < 1.82 \text{ rad/s}) \\
\beta_r &= 1.9969\Omega - 35.681 \quad (\Omega \geq 1.82 \text{ rad/s})
\end{align*}
\]  
(34)
(35)

The other variables satisfy the following inequalities.

\[
\begin{align*}
0.94 \text{ rad/s} &\leq \Omega \leq 2.84 \text{ rad/s} \\
3.00 \text{ m/s} &\leq V_1 \leq 38.26 \text{ m/s} \\
\beta_m &\leq \beta_2 \leq 90.0^\circ \\
b &\leq \gamma \leq 8.0^\circ \\
0.0^\circ &\leq \psi \leq 360.0^\circ
\end{align*}
\]  
(36)

(2) For the 2.0 MW VSVP wind turbine, the constraint conditions were

\[
\begin{align*}
\beta_m &= \max(\beta_r, \beta_v) \quad (V_1 \leq V_o) \\
\beta_m &= \beta_v \quad (V_1 > V_o)
\end{align*}
\]  
(37)

where \( \beta_v \) can be formulated as:

\[
\begin{align*}
\beta_v &= 0 \quad (V_1 \leq 10.5 \text{ m/s}) \\
\beta_v &= -0.0476(V_1)^2 + 2.9958V_1 - 22.756 - 8.0 \quad (10.5 \text{ m/s} < V_1 \leq 25 \text{ m/s})
\end{align*}
\]  
(38)
(39)

\[
\begin{align*}
\beta_v &= 22.389 - 8.0 + 5(V_1 - 25) \quad (V_1 \geq 25 \text{ m/s})
\end{align*}
\]  
(40)

and \( \beta_r \) is:

\[
\begin{align*}
\beta_r &= 0 \quad (\Omega < 1.81 \text{ rad/s}) \\
\beta_r &= 3.5594\Omega - 60.153 \quad (\Omega \geq 1.81 \text{ rad/s})
\end{align*}
\]  
(41)
(42)
The other variables can be defined as:

\[
\begin{align*}
0.94 \text{ rad/s} \leq \Omega &\leq 2.45 \text{ rad/s} \\
3.00 \text{ m/s} \leq V_1 &\leq 38.87 \text{ m/s} \\
\beta_m &\leq \beta_2 \leq 90.0^\circ \\
-8.0^\circ &\leq \gamma \leq 8.0^\circ \\
0.0^\circ &\leq \psi \leq 360.0^\circ
\end{align*}
\]

(43)

Then, in order to compare with the results from FOCUS, the load factor \( \gamma_F \) will be considered. According to the GL standard, \( \gamma_F \) is set to be 1.35.

Moreover, the related parameters in the improved PSO algorithm are prefixed as follows:

1. Number of individuals: 30;
2. Maximum number of iterations: 100;
3. Probability of selection \( P_s \): 0.0333;
4. Maximum inertial weight: 0.6253;
5. Minimum inertial weight: 0.0562

3.1. Extreme Loads at the Blade Root

Finally, the maximum flapwise bending moment of blade root in chord coordinate (CMF_ROOT) and the maximum edgewise bending moment of blade root in chord coordinate (CME_ROOT) was calculated respectively by the PSO-ELPM model. Each of them was computed ten times independently. Figures 8 and 9 give the iterative processes, from which the extreme CMF_ROOT and the extreme CME_ROOT was obtained. The extreme values of CMF_ROOT and CME_ROOT are shown in Table 1. The status parameters of the wind turbine, while the extreme loads were reached, are given to highlight in Table 2.

### Table 1. The extreme values of CMF_ROOT and CME_ROOT.

|            | 1.5 MW       | 2.0 MW       | Error | 1.5 MW       | 2.0 MW       | Error |
|------------|--------------|--------------|-------|--------------|--------------|-------|
| CMF_ROOT (kNm) | 4176.45     | 4197.57     | 0.51% | 8084.68     | 7928.81     | −1.93% |
| CME_ROOT (kNm) | 2571.43     | 2741.36     | 6.61% | 4542.79     | 4352.48     | −4.19% |

### Table 2. Extreme bending moments and their corresponding status parameters with considering the constraint conditions about the \( \beta_m \).

| Bending Moment (kNm) | \( V_1 \) (m/s) | \( \Omega \) (rad/s) | \( \beta_2 \) (\(^\circ\)) | \( \gamma \) (\(^\circ\)) | \( \psi \) (\(^\circ\)) |
|----------------------|----------------|--------------------|-----------------|-----------------|----------------|
| 1.5 MW               | CMF_ROOT       | 4197.56            | 25.00           | 2.46            | 11.81          | −1.94  |
|                      | CME_ROOT       | 2741.36            | 29.13           | 2.84            | 32.48          | −8.00  |
| 2.0 MW               | CMF_ROOT       | 7928.81            | 15.95           | 1.91            | 4.92           | −8.00  |
|                      | CME_ROOT       | 4352.48            | 38.87           | 2.45            | 90.00          | −8.00  |

From Table 1, it can be seen that the results using PSO-ELPM are well in agreement with those using FOCUS, with a difference of less than 10%. Especially for the extreme CMF_ROOT, the error was lower than 5%. It shows that PSO-ELPM can be used for predicting the extreme load. Moreover, PSO-ELPM needs much less computational cost than FOCUS. That is, PSO-ELPM only needs about 50 min to get the extreme CMF_ROOT (ten times calculation) while FOCUS needs about 12 h (one-time calculation). Hence, PSO-ELPM is very suitable for preliminary blade design, and even optimization design.

Furthermore, from Figures 8 and 9, it is clear that the status parameters of the wind turbine changed with iteration. When the maximum load was obtained, the corresponding values of these
parameters were immediately determined in PSO-ELPM. This is an advantage of PSO-ELPM while FOCUS and other statistical models cannot do it. This will be very useful to analyze the conditions while the extreme loads occur. If we can avoid this condition by control or other methods, the extreme loads will largely reduce.

Then, in order to compare with the results from FOCUS, the load factor $\gamma_F$ will be considered. According to the GL standard, $\gamma_F$ is set to be 1.35.

Moreover, the related parameters in the improved PSO algorithm are prefixed as follows:
1) Number of individuals: 30;
2) Maximum number of iterations: 100;
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**Figure 8.** Iterative processes of variables: (a) Extreme CMF_ROOT calculation; (b) Extreme CMF_ROOT calculation for the 1.5 MW wind turbine blade.

From Table 2, we can see that the values of $\gamma$ were all negative. This reveals that the negative yaw angle made it easy to generate the extreme bending moments. For the pitch angle, it is always close to the $\beta_r$, which indicates that smaller pitch angle will obtain larger values for the extreme bending moments. This can also be used to explain why the control rules always increase the pitch angle when the wind speed exceeds the rated value. For the rotor speed, the extreme CMF_ROOT does not get its upper bound while the extreme CME_ROOT does. The reason may be the function of the constraints, which limits the rotor speed corresponded to the pitch angle and the wind speed. To further certify the above speculation, the constraint conditions about the $\beta_m$ are not considered. In this case, $\beta_2$ will be decided by the mechanical property of the wind turbine and meet Equation (44).

$$\beta_u \geq \beta_2 \geq \beta_l$$  \hspace{1cm} (44)

Table 3 gives the extreme bending moments at the root without considering the constraint conditions about the $\beta_m$ and their corresponding status parameters of the wind turbine calculated by PSO-ELPM. We can see that $\Omega$ gets the upper bound at the extreme loads. These results indicate that the above speculation is correct.

From Tables 2 and 3, it also shows that larger rotor speed generated larger bending moments. Therefore, we need to brake the rotor when $\Omega$ exceeds the upper bound. In addition, the extreme CMF_ROOT largely increased without considering the constraint conditions about the $\beta_m$, while the extreme CME_ROOT was almost unchanged. This may be because $\Omega$ had larger influence on the
extreme CME_ROOT compared with the other variables. CME_ROOT of 2.0 MW gets the upper bound of the rotor speed while pitch angle reached the maximum. These results do not seem to represent the reality of the situation. However, it gives the limit state of the wind turbine, which may happen with the wind direction changing rapidly.

Table 3. Extreme bending moments and their corresponding status parameters without considering the constraint conditions about the \( \beta_m \).

| Bending Moment (kNm) | \( V_1 \) (m/s) | \( \Omega \) (rad/s) | \( \beta_2 \) (°) | \( \gamma \) (°) | \( \psi \) (°) |
|----------------------|-----------------|---------------------|-----------------|---------------|-------------|
| 1.5 MW CMF_ROOT       | 8328.10         | 38.26               | 2.84            | 3.50          | −8.00       | 25.90       |
| 1.5 MW CME_ROOT       | 2801.07         | 38.26               | 2.84            | 37.53         | −8.00       | 52.73       |
| 2.0 MW CMF_ROOT       | 17877.90        | 38.87               | 2.45            | 7.20          | −8.00       | 16.34       |
| 2.0 MW CME_ROOT       | 4352.48         | 38.87               | 2.45            | 90.00         | −8.00       | 339.97      |

3.2. Extreme Load Distribution of the Blade

Though the extreme loads at the root can be used to evaluate whether the blade is overloading for the wind turbine, as well as the safety of the blade bolts, it is not enough for the preliminary blade preliminary. The load distribution along the blade span is also important. Luckily, this model can obtain the extreme load at any section. To further test it, the extreme load distributions for the 1.5 MW blade and 2.0 MW blade were obtained and compared with FOCUS results respectively in Figures 10–13. The extreme loads were close at most sections for flapwise bending moments, though there was little difference in the middle of the blade in Figures 10 and 12. From Figures 11 and 13, we can see that the edgewise bending moment was also close at the blade tip, while there was obvious differences near the root. This method needs further improvements for edgewise load prediction. Overall, the PSO-ELPM model can accurately predict extreme bending moments in flapwise, which
has the main influence on blade design. Maybe these loads can be used for evaluating the maximum blade deflection, which is a very important constraint for blade design. Moreover, from Table 2, we can see the extreme bending moments in edgewise happen at the maximum rotor speed. However, it is not easily determined due to $\Delta \Omega$. Maybe this is the reason why these loads cannot be accurately evaluated.

![Graph](image_url)

**Figure 10.** Comparison of extreme flapwise bending moment along the blade (1.5 MW).

![Graph](image_url)

**Figure 11.** Comparison of extreme edgewise bending moment along the blade (1.5 MW).

### 3.3. Effects of $\Delta \Omega$

Based on the above analysis, $\Delta \Omega$ might have a significant effect on the extreme load prediction in this model. In this section, we want to test the results when $\Delta \Omega$ is equal to 0, that is without delay in the safety control system. During the calculation, Equations (20) to (33) were still used. However, the upper value of rotor speed was change to 2.45 rad/s in Equation (26) and 2.17 rad/s in Equation (33) respectively. Figures 14–17 show the comparison of results from PSO-ELPM with and without considering $\Delta \Omega$ (expressed as “no delta omega” in these figures).
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From Figures 14 and 16, we can see the flapwise bending moments are still very close. It means $\Delta \omega$ had little influence on them. This is also certified by Table 2. In Table 2, the extreme flapwise bending moment did not happen at the maximum rotor speed. Hence, the extreme value was not changed though the maximum rotor speed reduced. But the edgewise bending moments were very different, and noticeably changed without $\Delta \omega$, especially near the blade root. Therefore, it needs further improvement. The $\Delta \omega$ should be fixed to make this model more accurate.

Figure 12. Comparison of extreme flapwise bending moment along the blade (2.0 MW).

Figure 13. Comparison of extreme edgewise bending moment along the blade (2.0 MW).

Figure 14. Comparison of extreme flapwise bending moment along the blade (1.5 MW).
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Figure 15. Comparison of extreme edgewise bending moment along the blade (1.5 MW).

Figure 16. Comparison of extreme flapwise bending moment along the blade (2.0 MW).

Figure 17. Comparison of extreme edgewise bending moment along the blade (2.0 MW).

4. Conclusions

In this paper, so as to quickly solve the extreme loads and make it easily usable for preliminary blade design, a new simplified method, called as PSO-ELPM, was developed and validated in this study. Following this, the effects of the status parameters on the extreme loads were analyzed. Some conclusions are given below:

1. The extreme root loads computed by PSO-ELPM and FOCUS were very close. The error between them was less than 10%. The extreme CMF_ROOT was lower than 5%. Moreover, PSO-ELPM needs much less computation cost than FOCUS. Hence, PSO-ELPM can be used for predicting the extreme load and is suitable for preliminary blade design.
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1. The extreme root loads computed by PSO-ELPM and FOCUS were very close. The error between them was less than 10%. The extreme CMF_ROOT was lower than 5%. Moreover, PSO-ELPM needs much less computation cost than FOCUS. Hence, PSO-ELPM can be used for predicting the extreme load and is suitable for preliminary blade design.

2. Higher rotor speed and smaller pitch angle will generate larger extreme bending moments at the root. It is for this reason that we need to pitch the blade to feather at high wind speed and activate security strategies when the rotor speed is larger than overspeed. In addition, negative yaw angle easily generates the extreme bending moments.

3. When the control system is inactive, the extreme CMF_ROOT is increased but the extreme CME_ROOT is almost unchanged. It shows that the control system has significant influence on reducing the extreme loads, especially the extreme CMF_ROOT.

4. By comparison of the extreme load distribution, flapwise bending moments are close to the results of FOCUS, while edgewise bending moments need be improved in the future.

5. By the analysis of the effects of \( \Delta \Omega \), there is little influence on the extreme flapwise bending moments. However, the edgewise bending moments change a lot while the maximum rotor speed reduces.

The PSO-ELPM model is a new method, which gives a different way to solve the extreme loads exerted on wind turbine blades. It has some advantages—it is simple and fast, but there are some aspects which have not been considered. There are also some defects that need to be further improved. For example, some parameters that change with the wind turbine model and not easily determined, and it needs to be fixed to make this model more generic. A further example is that the load calculation model needs to be improved so as to obtain better results. For example, the \( \Delta \beta_2 \) should be set as a variable changing with the wind speed. Storm conditions can be the most challenging conditions and should be added in the improved model. The load calculation could consider the aeroelastic effect. The \( \Delta \Omega \) should be fixed to make this model more accurate.

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