



## Smart Variable Rotor of Vertical Axis Wind Turbine with Faster Cut-in Speed and Wide Range Extract Power Using Artificial Intelligent

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**Abstract:** This paper presents the standalone type vertical axis wind turbine (VAWT) smart rotor using variable diameter rotor (VDR) in order to tap constant power and maintain cut-in wind speed. VDR is a smart variable rotor capable of operating at a low wind speed, in which the width of diameter rotor is adjustable using actuators. The VDR rotor is connected to the permanent magnet synchronous generator (PMSG) and a DC-DC boost converter. The controller of VDR rotors uses fuzzy logic controller (FLC). The FLC variable inputs are wind speed data and power output, broken down into cluster groups to determine the diameter rotor position. The wind speed data as fuzzy input are produced by wind speed estimator using artificial neural network (ANN) to maintain cut-in speed to be faster. The velocity movement of VDR is limited from 75 cm to 150 cm. The VDR extension increases by 25% when the wind speed decreases from 8 to 6 m/s to obtain consistent power at 150 Watt. The experimental of VDR system is tested at low wind speeds ranging from 1 to 8 m/s as a verification of the control system. The result showed that the VDR produced five times increase in efficiency with faster cut-in wind speed at 2.0 m/s.

**Keywords:** Smart rotor, Variable speed, Variable diameter, Swept area, Dynamic rotor.

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### 1. Introduction

Rapid occurrence of global warming and decreasing energy reserves has become a global issue and a challenge for researchers. One of many means of preservation is to develop renewable energy sources that are environmentally friendly, such as solar cell clustering, wind farm optimization, and wind turbine control optimization by incorporating modern technology to improve the efficiency of renewable energy [1-5].

Generally, wind turbines are divided into two types based on the rotation, namely horizontal axis wind turbines (HAWTs) and vertical axis wind turbines (VAWTs). HAWTs are usually used on a large scale wind turbine power, but these turbines are highly wind dependent, require a complex yaw mechanism, are very expensive [6, 7]. On the other hand, VAWTs have much better performance and are suitable in rural and remote areas because they are able to operate in all wind directions, and the

generator can be placed close to the ground which, in turn, results in easier maintenance and lower cost [8, 9].

The power output of wind turbine depends on wind speed, power coefficient, and swept area. The swept area as rotor of VAWTs consists of blades and rotor diameter. The rotor blades with variable pitch control are usually used for aerodynamic optimization on the side of the rotor using ANN, fuzzy, PID and other artificial intelligence controls to limit the power output during strong wind conditions or above the wind speed rating [10-14].

Smart rotor design for aerodynamic optimization with the variable swept area (VSA) concept or telescopic blade is designed to achieve constant power [15-19]. The controls used are fuzzy, PID by applying adjustable swept area that the turbine is capable of operating at low wind speeds or being activated at either below wind speed rating or above wind speed rating. When below the wind speed rating,

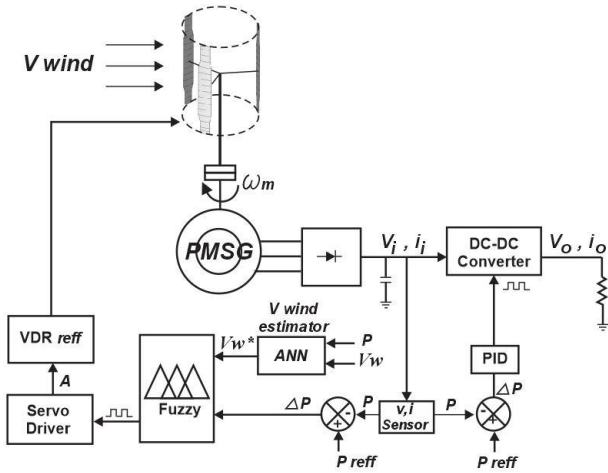


Figure. 1 VAWT control system with VDR

the VSA size will be maximum; when above the wind speed rating, the VSA will track the constant power by changing the VSA size close to the minimum position. Size and dynamic rotor with blade transitioning capability variables aims to increase the power coefficient of VAWT [20].

However, the aforementioned controls cannot maintain cut-in speed when the wind speed is at below rating. In a case when the wind speed is below its rating speed, the variable pitch control will remain at optimum fixed pitch angle [21-23], pitch angle change occurs only when the turbine operation passes the wind speed rating to limit the power output. Meanwhile, variable swept area at low wind speed will remind at maximum position.

The contribution of this paper is to accelerate cut-in speed and maintain the constant power of VAWTs using VDR as smart rotor as shown in Fig. 1. VDR has an adjustable rotor diameter capable of extending and retracting based on wind speed using FLC and ANN on rotor side, and PID controller on converter side. The VDR rotor is connected to permanent magnet synchronous generator (PMSG) and a DC-DC boost converter with PID controller. FLC is applied on the actuator interface of VDR, with FLC input being data on wind speeds and generator power output. Wind speed data are produced by wind speed estimator using ANN. By determining the wind speed earlier, the system can work better to accelerate the cut-in speed, so that short term wind speed forecasting is needed using an ANN placed as FLC input. The performance of the proposed method is evaluated both by simulation and laboratory experiments.

This paper is organized as follows. Section 2 presents the VAWT modeling concept. Section 3 discusses the proposed VDR system. Section 4 explains the experimental set-up. The assessment of

the proposed method is presented in Section 5. Finally, section 6 concluded this paper.

## 2. Proposed VDR system

### 2.1 Operating cut-in speed region

The relationship between wind speed and generator power is divided into three categories as shown in Fig. 2. Cut-in wind speed is the lowest wind speed to start producing power, rated wind speed is the minimum wind speed to procure rated power, and the maximum speed over rated power is denoted as cut-out wind speed. In region 1, the wind speed is lower than cut-in, in which general of wind turbine fails to produce power output due to low wind speed being below the cut-in wind speed. The region 2 position is between cut-in and rated wind speed, in which the wind captured is transformed into valuable power. Region 3 is in between rated wind speed and cut-out wind speed as full load region where a constant power is realized [24, 25].

In VDR, the controller is designed into three operating regions. The first one is below cut-in region; the turbine works at lower  $C_p$  to increase the power at lower wind speed, the position of the rotor diameter is adjusted into optimum diameter with cut-in wind speed between 1.5-2.5 m/s by changing the diameter size. Changes in rotor diameter can be seen in the VDR concept features in Fig. 3 and Fig. 4.

The optimum reference diameter at the lowest wind speed will be the membership function of FLC output to tap cut-in wind speed in which the lowest wind speed data must be identified earlier using short-term forecasting with ANN. When passing the rated power region limit, the control strategy is to maintain the rated power constant by reducing the diameter size and minimizing aerodynamic loads of the wind turbine.

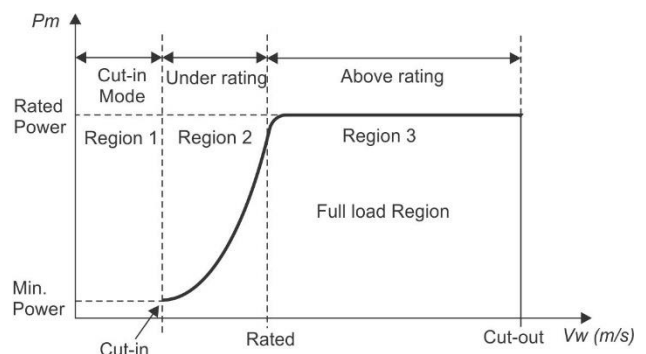


Figure. 2 VAWT control system with VDR

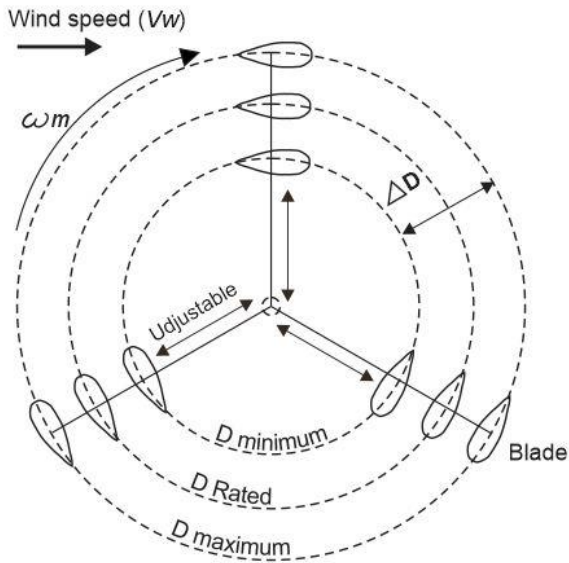


Figure. 3 The VDR rotor concept

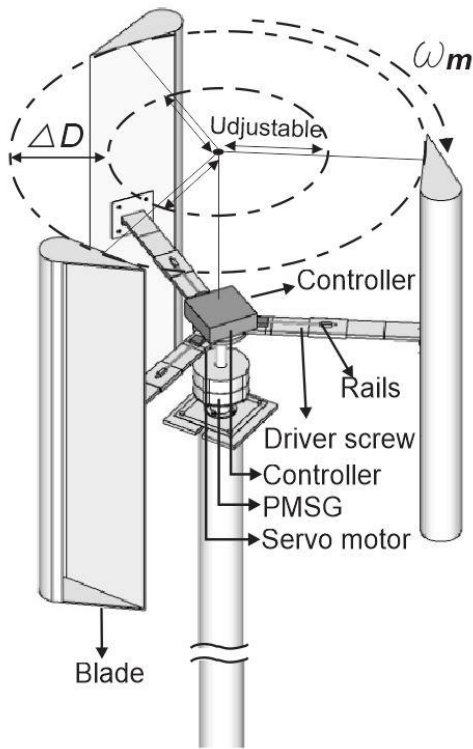


Figure. 4 The VDR feature

**2.2 VAWT model**

The concept of wind energy converting mechanical power into electrical energy is presented in Fig. 5, in which parameter  $D$  is the rotor diameter set to be adjustable. The wind power captured by the rotor turbine/blades is converted into mechanical power calculated as follows:

$$P_m = \frac{1}{2} \rho \cdot A \cdot v^3 \cdot C_p(\theta, \lambda) \tag{1}$$

In which  $C_p(\theta, \lambda)$  is a power coefficient of the blade as a nonlinear function of tip speed ratio ( $\lambda$ ) and pitch angle ( $\theta$ ). The maximum  $C_p$  value of the current advanced technology according to the Betz limit theory is 0.59 [26].  $A$  is swept area ( $m^2$ ) which captures  $v$  as cubic function of wind speed ( $m/s$ ), and  $\rho$  is air density of  $1.225 \text{ kg/m}^3$ .

Swept area  $A$  consists of turbine height  $h$  and rotor diameter  $D$ , in which diameter  $D$  in design variables, ranging from 75 cm to 100 cm, which means that diameter variable is a diameter capable to extend and retract to face dynamic wind speed as shown Fig. 3. This diameter change is controlled by FLC on the radius side of the turbine which has three blades as VDR, in which each change in the variable diameter causes the swept area to change, so the power output also changes. It is affected by the tip speed ratio (TSR) and  $C_p$  during the transition change of the diameter. TSR is defined as the ratio of the blade tip speed to the incoming wind speed, in which wind turbine can achieve wider range extract power by maintaining optimal TSR at different wind speed. The TSR  $\lambda$  is formulated as follows [27].

$$\lambda = \frac{\omega_m r_t}{v_w} \tag{2}$$

Where  $\lambda$  is the TSR,  $\omega_m$  is the angular speed of turbine ( $rad/s$ ),  $r_t$  is the radius of turbine and  $v_w$  is the wind speed ( $m/s$ ). The mechanical angular speed of the rotor and torque can be formulated as:

$$J \frac{d\omega_m}{dt} = T_e - T_m - B \cdot \omega_m \tag{3}$$

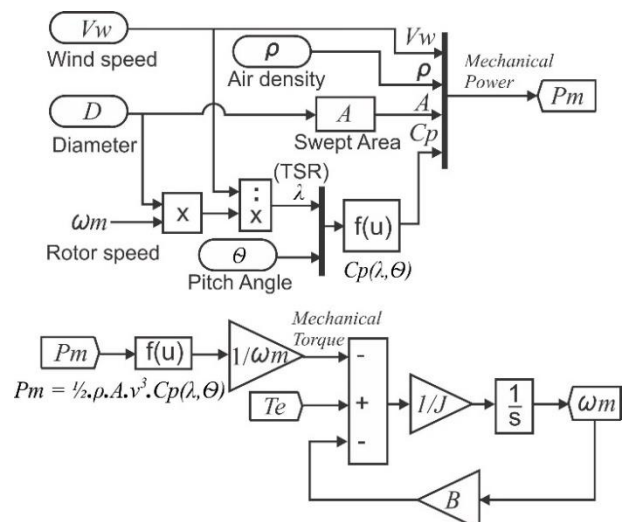


Figure. 5 Wind turbine model

Where  $J$  is the combined inertia of rotor and the wind turbine ( $kg.m^2$ ),  $T_e$  is the electromagnetic torque ( $N.m$ ), and  $B$  is the viscous friction of the rotor ( $N.m.s/rad$ ). Based on Eq. (1) and Eq. (2), the mechanical torque of wind turbine can be formulated as:

$$T_m = \frac{P_m}{\omega_m} = \frac{\frac{1}{2} \rho \cdot A \cdot v^3 \cdot Cp(\theta, \lambda)}{\omega_m} \quad (4)$$

Where  $T_m$  is the mechanical torque of the turbine ( $N.m$ ). To optimize the mechanical efficiency at different wind speeds, there is only one appropriate rotor speed with maximum  $Cp$  value and optimum tip speed ratio [28,29].

### 2.3 The VDR as smart rotor

The VDR has a variable diameter rotor which can be retracted and extended based on wind speed conditions predicted by ANN as wind speed estimator. Changes on VDR size are controlled using FLC; the fuzzy inputs are wind speed and power turbine and the fuzzy output is diameter reference. VDR will affect the power output at the turbine as shown by Eqs. (5) and (6), in which  $A$  is the swept area,  $h$  is the height of blade, and  $D$  is the diameter rotor.

$$A = h \cdot D \quad (5)$$

Then the VDR can be calculated as follows:

$$A = h(D + \Delta D) \quad (6)$$

Where  $\Delta D$  is the difference between maximum and minimum position of diameter configuration as shown in Fig. 4. Fully extended is the maximum rotor diameter when moving outward, while fully retracted is the minimum retraction of the rotor diameter when being retracted. The calculation of power produced by VDR is shown below:

Table 1. VAWT parameter

Parameter	Value
Rated mechanical power	500 Watt
Number of Blade	3
Diameter max	100 cm
Diameter min	75 cm
Blade height	100 cm
Blade type	Lenz

Table 2. Fuzzy input

	Wind Speed (m/s)	Power (W)	Fuzzy State
Cluster 1	[1 – 4.68]	[0 – 312.5]	Cut-in speed
Cluster 2	[1.87 – 7.5]	[125 – 500]	Low speed
Cluster 3	[4.68 – 10.31]	[312.5 – 687.5]	Rating speed
Cluster 4	[7.5 – 13.12]	[500 – 875]	Strong
Cluster 5	[10.31 – 15]	[687.5 – 1000]	Very Strong

$$Pm = \frac{1}{2} \cdot \rho \cdot (h \cdot (D + \Delta D)) \cdot v^3 \cdot Cp(\theta, \lambda) \quad (7)$$

The operating region of generator is when the wind speed reaches its rating. VDR will be adjusted to obtain the appropriate generator power output. VDR will be retracted when wind speed increases above the generator rating power. The mechanism of VDR system when extending or retracting based on particular wind speed becomes the fuzzy logic rule base.

The movement of VDR is adjusted by wind speed clustering as membership function input of Fuzzy Logic Controller. The linguistic variables are used to substitute the mathematical model of the dynamic system [30,31].

The VDR block diagram with FLC and ANN can be seen in Fig. 6. In the experiment, there are two ANN input signals, namely P and Vw, P is the analog power output signal from the current and voltage sensors. It is then converted into 8 bit digital data via the ADC input on the microcontroller. Vw is a wind speed sensor, or anemometer in the form of an 8-bit digital signal ranging from 1 m/s - 12 m/s. Both P and Vw states are obtained from ANN offline training which is performed in a microcontroller. It yields Vw\* state as wind speed forecast. This state is a digital signal from wind speed forecasting that becomes as an input for Fuzzy. Meanwhile, there are signals P and P ref, signal P is an analog signal from the power output P converted into a digital signal through the ADC. P ref is the digital reference value to obtain the error value of P and  $\delta eP$  using the following equation.

$$eP(k) = P(k) - P_{ref}(k) \quad (8)$$

$$\delta eP = eP(k) - eP(k-1) \quad (9)$$

The power output  $\delta eP$  is obtained from recent value  $eP$  minus with the previous error, with  $Z^{-1}$  being the sampling time delay. The output Fuzzy is a digital

signal to drive the servo motor and produced VDR reference.

The process of fuzzification is assigned to the linguistic variables. The wind speed cluster as verbal value is divided into five fuzzy subsets, namely cut-in speed (CS), low speed (LS), rating speed (RS), strong (S) and very strong (VS) as shown in Fig. 7. The membership function used is the triangle function.

The fuzzification process uses the center gravity method to transform the fuzzy output into the crisp output [31,32]. The adjustment of the rotor diameter position and variable wind speed from the sensor is processed with 25 number “if-then” rules, in which the blurring rule base is processed by inferencing system using the Mamdani method. The rules of the VDR reference is determined by considering fuzzy rules as follows:

Rules  $i$ : If  $V_w$  is  $X_i$  and  $eP$  is  $Y_i$  and  $\delta eP$  is  $Z_i$ ,

Then  $VDR_{ref}$  is  $F_i; i=1, \dots, m$ .

Where  $X_i, Y_i, Z_i$  are the fuzzy set and  $F_i$  is fuzzy singleton.

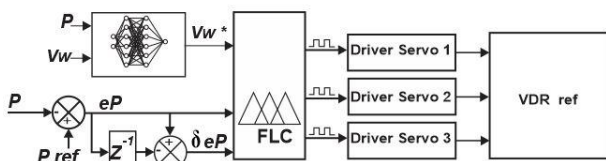


Figure. 6 The VDR control block diagram

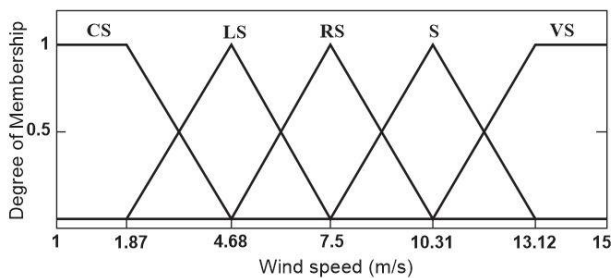


Figure. 7 Membership function of wind speed

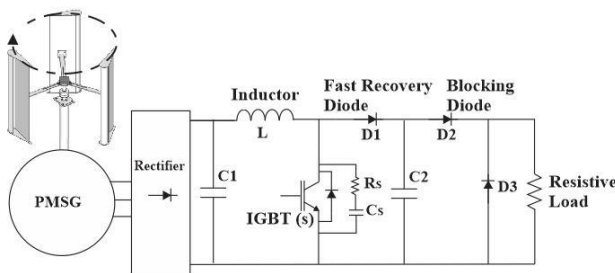


Figure. 8 Boost converter block diagram

Table 3. Parameter of boost converter

Parameter	Value
Input Voltage	35 Volt
Output Voltage	70 Volt
Boost Inductor	92 $\mu$ H
Filter Capacitor	378 $\mu$ F
Resistive Load	25 ohm
Kp	0.001
Ki	0.01

Fig. 8 shows that VAWT is connected to the power converter on its PMSG to tap constant voltage. The power converter consists of an unregulated diode rectifier, which transforms AC to DC voltage [33]. The output of boost converter is connected to the resistive load in which the duty cycle configuration is controlled by the PID controller to generate pulses on IGBT(s). The ratio of duty cycle can be calculated as follows [34]:

$$D_{cl} = \frac{T_{on}}{T_s} 100 \% \quad (10)$$

The voltage output of boost converter is greater than the DC voltage input because the voltage output is the sum of voltage input and inductor voltage [35-37]. The inductor voltage in period ( $T_s$ ) is zero and depends on the duty cycle that can be formulated as:

$$D_{cl} \cdot V_{dc} + (1-D_{cl})(V_{dc} - V_{out}) = 0 \quad (11)$$

The input and output voltage can be formulated as:

$$V_{out} = \frac{V_{in}}{1 - D_{cl}} \quad (12)$$

Where  $D_{cl}$  is duty cycle of the converter, defined by  $D_{cl} = T_{on}/T_s$ ;  $T_s$  is the switching period;  $T_{on}$  and  $T_{off}$  are the turn-on and turn-off times of the switch  $S$ , respectively. Eq. (12) indicates that the  $V_{out}$  of the converter is always higher than its input voltage. Where Eqs. (11) and (12) are for the continuous conduction mode (CCM) of the boost converter, assuming the inductor  $L$  is pure inductor. CCM mode is used because to get a low output voltage ripple on the stand-alone load.

Power input equals to the power output by neglecting the losses of power converter.

$$P_{in} = \frac{V_{dc}^2}{R_{in}} \cong P_{out} = \frac{V_{dc}^2}{(1-D_{cl})^2 R_L} \tag{13}$$

To tap constant power, PID controller is applied to the DC-DC boost converter of which equation is as follows:

$$u(t) = k_s e(t) + K_i \int_0^t e(t) + K_d \frac{de(t)}{dt} \tag{14}$$

$$C(s) = k_p + \frac{K_i}{S} + K_d S \tag{15}$$

Where (K<sub>i</sub>) and (K<sub>d</sub>) are integration gain and derivative gain of each constant.

### 3. Wind speed estimator with ANN

ANN is particularly used to implement nonlinear time varying input-output mapping that develops intelligent information processing system, is highly parallel and good fault-tolerance [38]. ANN is trained using data sampling on the behavior of a system. The ANN in this study used multilayer perceptron (MLP) for wind speed forecasting with off-line training. The purpose of VDR using an ANN is to establish the intelligent control of rotor diameter based on the information extraction of the wind power characteristics.

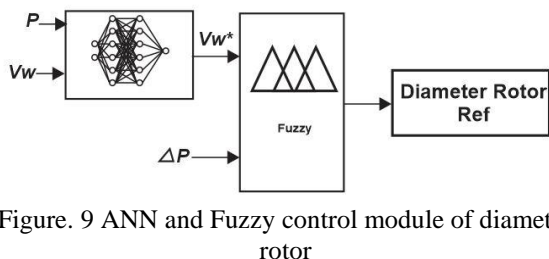


Figure. 9 ANN and Fuzzy control module of diameter rotor

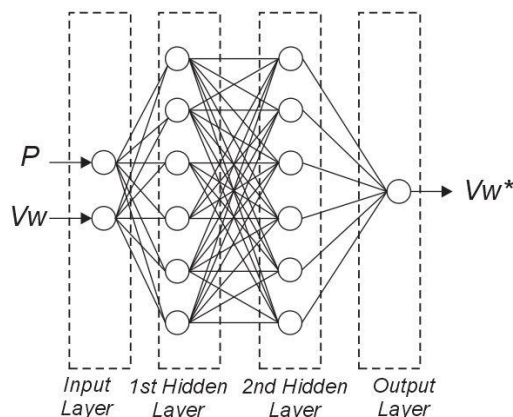


Figure. 10 Wind speed prediction ANN-based

This method requires exact past wind speed data using wind speed sensor in which the wind speed prediction result or the ANN output becomes Fuzzy input to determine the correct diameter reference as shown Fig. 9. The proposed forecasting model based on ANN has two inputs and one output for short time prediction of wind speed as shown in Fig. 10. The inputs are generator output (*P*) and wind speed (*V<sub>w</sub>*) data samples, recombined as “n” data pairs (*P*<sub>(i)</sub>, *V<sub>w</sub>*<sub>(i)</sub> | *i* = 1~n) which become the matrix input vector of ANN.

The wind speed data samples are used as target to train the layer networks, consisting of five tan sigmoid neurons in hidden layer and one linear output layer. The forecasting performance were tested using wind speed data which had never been presented to the ANN.

The wind speed forecasting with off-line training generated an average wind speed range of 1 m/s – 8 m/s, the sampling period is 1 second with 500 sample data pairs. The test result is shown in Fig. 11 with an average error of 0.15 in which ANN can better track the test samples and be used to predict velocity.

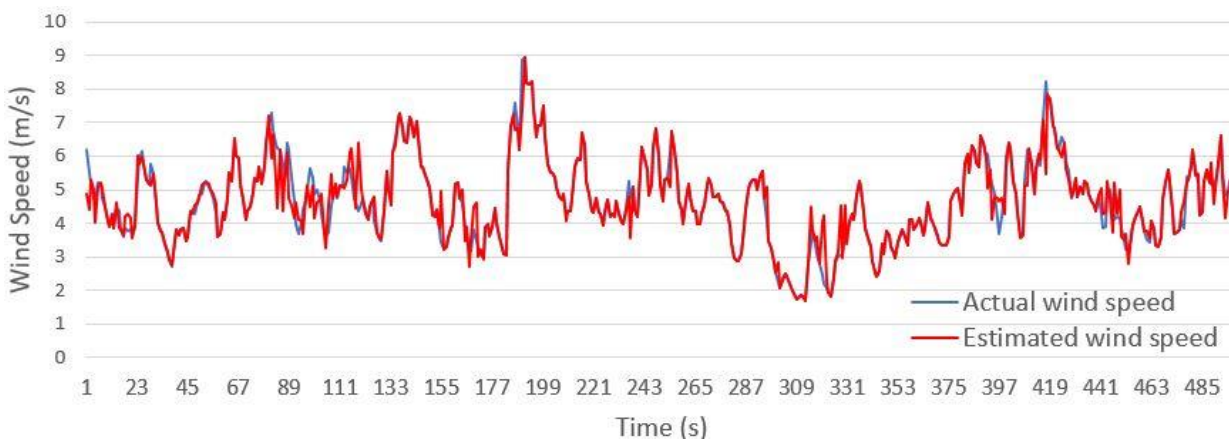


Figure. 11 Wind speed prediction

### 4. Experimental of prototype

The VDR system configuration consists of a fan, VDR rotor, PMSG, converter, and resistive load as shown in Fig. 12. The input variable wind speed used a fan of which wind speed ranged from 1 m/s to 20 m/s.

Fig. 13 shows the rotor diameter performance obtained for variable wind speed and different rotation (Rpm). On each diameter configuration, different power output characteristics are identified, in which these characteristics will be defined to select an operating point for the controller to work on. The optimum configuration of VDR to maintain cut-in speed is adjusted by wind speed clustering as Fuzzy membership function.

The diameter actuator obtains signal data from power controller to take appropriate power and track the turbine diameter. The actuator controller uses microcontroller to apply Fuzzy and stimulate the servo motors to adjust the diameter actuator movement. The adjustment of diameter based on wind speed forecasting uses ANN as FLC input. The FLC output is in the form of VDR reference to tap constant power.

VDR Rotor consists of variable diameter as seen in Fig. 14, in which variable diameter can extend its length at low wind speeds and retract at high wind speeds. The extension and retraction mechanism of variable diameter uses servo motor as shown Fig. 15. The VDR rotor hub is mounted to the shaft PMSG rotor as, with the PMSG output is connected to the three-phase rectifier, capacitor, converter and resistive load as shown in Fig. 8. The communication data of rotor controller and power controller uses wireless data interface as shown Fig. 16.



Figure. 12 The arrangement of main equipment

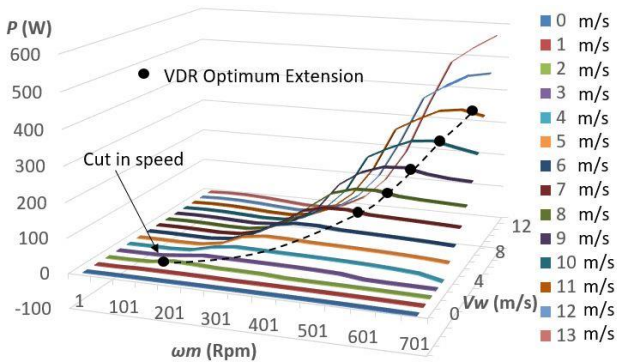


Figure. 13 Power curve of VDR extension

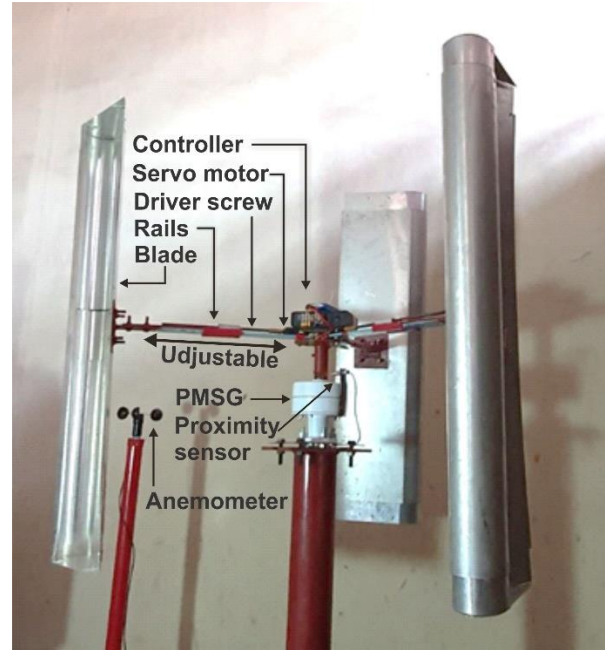


Figure. 14 The VDR actual photograph



Figure. 15 Variable diameter rotor assembly



Figure. 16 Instruments support and data logger

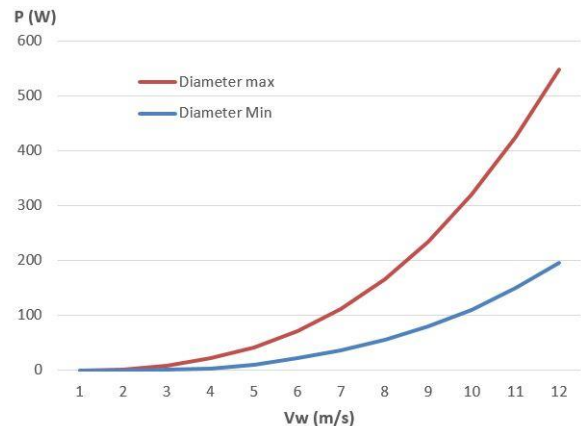


Figure. 17 The VDR power output

Fig. 17 shows that the power output is exponential to wind speed. The wider the diameter, the higher the power output. When wind speed increases the rotor speed, the VAWT power output increases at fully extended of diameter; but when the diameter is retracted, the power output decreases at the same range of wind speed variation. It means that changes in diameter can provide cut-in speed at low wind speeds from starting operation and provide constant power output at different variable wind speeds through modification of diameter to match with the wind speed.

## 5. Result and discussion

The VDR with ANN-Fuzzy PID was tested using simulation and experiment in the laboratory scale. The VDR system consists of a fan, VDR Rotor, PMSG, converter, and resistive load. A fan is used as a simulated wind speed input with various controlled speed ranges from 1 to 12 m/s. The VDR rotor is a straight blade type (aluminium) with a blade height of 100 cm, a diameter ranging from 75 cm to 150 cm, for which the diameter is adjusted using a servo motor as an actuator as shown in Fig. 14. The actuator control interface uses a microcontroller which has 54 input output pins, the crystal oscillator used is 16 Mhz. The VDR rotor hub is connected to PMSG 500 Watt. PMSG output is connected to the three phase diode bridge rectifier, thereby converting the AC voltage of PMSG output into DC voltage. The rectifier output current is measured using the ACS754 current sensor, while the rectifier output voltage is measured by the voltage divider circuit. Both of analog signals are sent to the analog-to-digital conversion in microcontroller. These measured current and voltage are then used for PID controller to determine PWM's duty cycle. This PWM signals are employed for switching the IGBTs boost converter through an opto-transistor circuit. The signal power output and wind speed data are processed by Fuzzy interface in microcontroller to drive the VDR actuator. Data logger is used to record the wind speed, current, voltage, and power measurement data to be presented in graphical form. The relationship between wind speed and PMSG power is obtained by the measurement of the PMSG VAWT performed in a laboratory.

The simulation in Fig. 18 shows the effectiveness of the proposed method with regards:

- The capability to maintain constant power during wind speed changes.
- Presenting better cut-in speed.

In which the proposed system has been compared with another papers [16, 29], in which the first paper uses Fuzzy-PID for variable swept area (VSA)

control [16], while the second paper uses the MLP ANN method for variable pitch control (VPC) [29]. On the other hand, the proposed system uses VDR with ANN-Fuzzy-PID. For fair comparison, the swept area of VDR, VSA, and VPC are using the same value and type (with small scale VAWT); moreover, three of which were tested at the same wind speed from 1 m/s to 12 m/s. Therefore, the three methods can be observed at any difference or changes in wind speed.

Fig. 18 shows the simulation measurement with wind speeds ranging from 1 m/s to 12 m/s. The result shows that VDR is able to start operation with faster cut-in wind speed of 2.2 m/s and maintain constant power at low wind speeds, compared to fuzzy-PID controller. When the wind speed decreases from 12 m/s to 10 m/s, VDR with ANN-Fuzzy-PID responds to the wind speed and obtains the correct VDR position varying the wind speed, and produces constant power while the wind speed changes. Moreover, VSA with Fuzzy-PID generates power output proportional to the wind speed; it can be seen at 0.7s when the wind speed decreases from 12 m/s to 10 m/s, the fuzzy PID power output drops from 200 Watt to 120 Watt, while at the same time the ANN-Fuzzy-PID holds constant power output at 200 Watt because the ANN predict the wind speed and becomes the input of Fuzzy controller to obtain the correct VDR position, thus accelerating the operation with faster cut-in wind speed at 2.2 m/s, compared to Fuzzy-PID, it can be seen at 0.1s in which the operation of Fuzzy PID starts later and is delayed 0.05 s at 0.15s with cut-in wind speed of 4.8 m/s.

VPC generates power output proportional to the wind speed. When the wind speed starts from 1 m/s to 12 m/s, the VPC cut in speed is 5.9 m/s, slower than VDR and VSA. When the wind speed decreases from 12 m/s to 10 m/s at 0.7 second, the VPC power output drops from 200 Watt to 108 Watt.

The constant power starts to re-increase when the wind speed is at 12 m/s at 0.8 second. It occurs because the pitch angle of the VPC is at the optimum angle position at low wind speed. Therefore, the power output is proportional to the wind speed, the change in pitch angle only occurs when the wind speed exceeds the rated value for the limiting power output. The experimental measurement can be seen in Fig. 19. The performance of VDR with ANN-Fuzzy-PID generates constant power despite changing wind speeds. It has been tested with random wind speed between 1 m/s to 8 m/s, in which the VDR starts the operation with cut-in wind speed at 2 m/s. The wind speed limit is determined based on ANN forecasting, in which the wind speed prediction becomes fuzzy input, and the fuzzy output is the



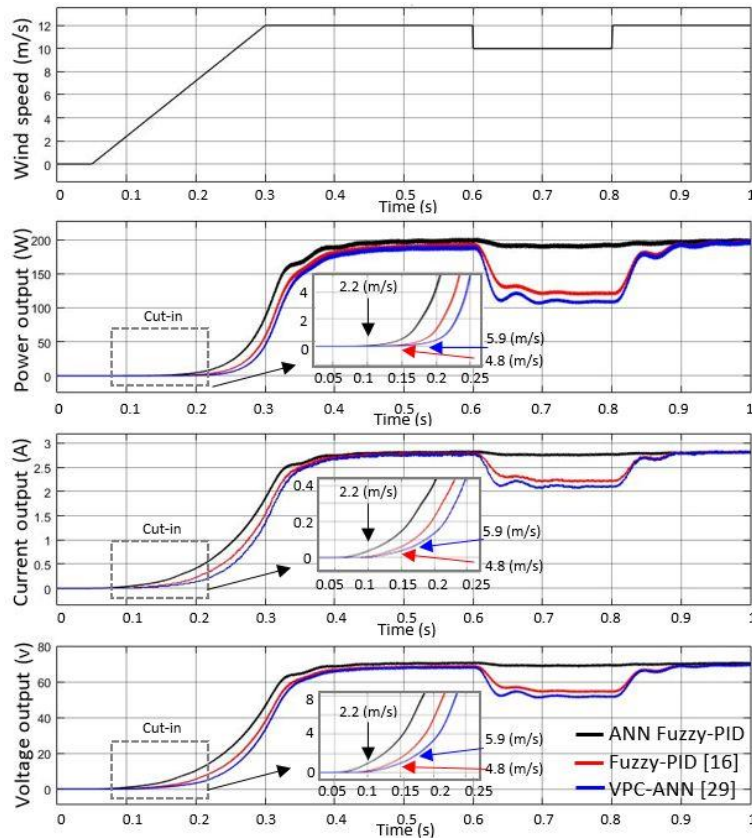
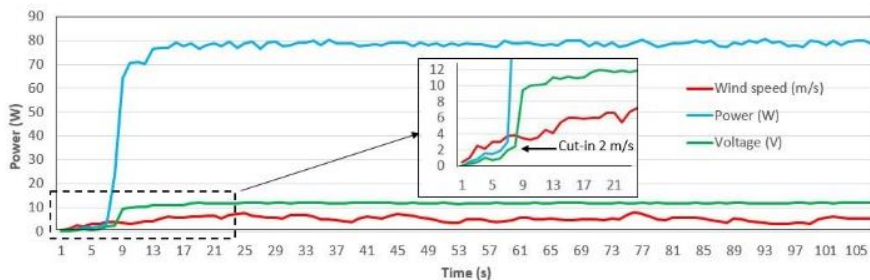
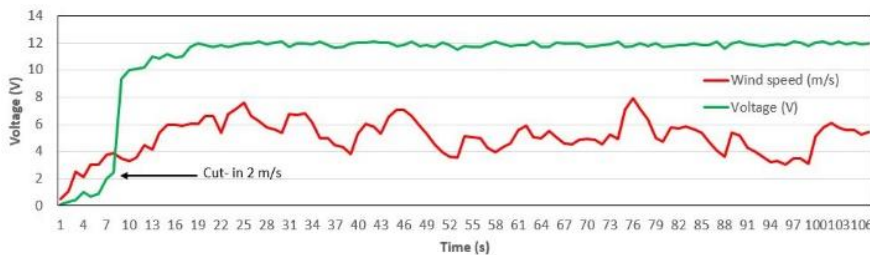


Figure. 18 Comparison performance of the proposed method VDR vs VSA [16] vs VPC [29]



(a)



(b)

Figure. 19 Performance of VDR with ANN-Fuzzy-PID: (a) power vs voltage vs wind speed and (b) voltage vs wind speed

diameter reference. The wind speed zone from cut-in speed to cut-out speed is divided into five cluster of wind speed as fuzzy membership function. The result shows that the VDR with ANN-Fuzzy-PID controller is far more powerful and generates better performance.

## 6. Conclusion

The VAWT performance with VDR method has been tested using simulation and experiment. It has been established that VDR with ANN-Fuzzy-PID is able to maintain constant power in spite of changing wind speeds, and optimization of VAWT starts at low

cut-in wind speed of 2.2 m/s. The VDR with ANN-Fuzzy-PID controller has better performance and manages the VDR actuator to overcome the wind speed and maintain constant power with an increased power by 37%. The result of VDR system with ANN-Fuzzy PID controller indicates that it is able to operate VAWT with faster cut-in wind speed and wider extract power at low wind speeds. Future research projects focusing on this topic should apply VDR with other artificial intelligent controller to increase the VAWT performance more efficiently.

### Conflicts of Interest

The authors declare no conflict of interests.

### Author Contributions

Planning and converter simulation, M.A.; VAWT experimenting, J.F.; wind speed forecasting verification, J.F.; model development, J.F.; data analysis, S.; writing the manuscript, M.A. and J.F.; theoretical framework development, J.F. and S.; supervision, M.A.

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### Notations

$A$	swept area
$B$	viscous friction of the rotor
$C_p$	power coefficient
$D$	diameter
$D_{cl}$	duty cycle
$h$	height
$J$	inertia
$K_p$	proportional gain
$K_i$	integral gain
$K_d$	derivative gain
$L$	inductor
$P$	power
$P_m$	mechanical power
$P_{in}$	power input
$P_{out}$	power output
$P_{ref}$	reference power
$R_{in}$	resistance
$R_L$	resistive load
$r_t$	radius
$T_e$	electromagnetic torque

$T_m$	mechanical torque
$T_{on}$	time on
$T_{off}$	time off
$T_s$	switching period
$t$	time
$V_i$	voltage input
$V_o$	voltage output
$V_{dc}$	DC voltage
$V_w$	wind speed
$\rho$	air density
$\theta$	pitch angle
$\lambda$	tip speed ratio
$\omega_m$	rotational speed
$\Delta D$	delta diameter
$eP$	error power
$\delta eP$	delta error power

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