Scaling effects of **rigid wing <u>fixed-wing</u>** ground-generation airborne wind energy systems

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Abstract. While some <u>Aa</u>irborne <u>Wwind Eenergy Ssystem</u> (AWES) companies aim at small, temporary or remote off-grid markets, others aim at utility-scale, multi-megawatt integration into the electricity grid. This study investigates the scaling effects of single-wing, ground-generation <u>Airborne Wind Energy Systems <u>AWES</u></u> from small to utility-scale systems, subject to realistic 10-minute, onshore and offshore wind conditions derived from the <u>a</u> numerical mesoscale weather research and

- 5 forecasting (WRF) model. To reduce computational cost, <u>vertical</u> wind velocity profiles are grouped into k = 10 clusters using k-means clustering. Three representative profiles from each cluster are implemented into a nonlinear AWES optimal control model, to determine power-optimal trajectories. We compare the effects of three different aircraft masses and two sets of nonlinear aerodynamic coefficients for aircraft with wing areas ranging from 10 m^2 to 150 m^2 , on operating parameters and <u>flight</u> trajectories. We predict size- and <u>weightmass</u>-dependent AWES power curves, annual energy production (AEP) and
- 10 capacity factor (cf) and compare them to a quasi-steady state reference model. Instantaneous force, tether <u>reeling</u> speed and power fluctuations as well as power losses associated with tether drag and system mass are quantified.

1 Introduction

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Airborne wind energy systems (AWESs) harvest wind energy from the stronger and less turbulent winds at mid-altitude, here defined as heights above 100 m and below 1500 m. These beneficial conditions promise more reliable and stable wind power

- 15 generation compared to the conventional wind turbines (WTs) at lower altitudes. The light, tower-less design allows for mobile deployment and reduces the capital cost of AWESs (Lunney et al., 2017). These kite_inspired systems consist of one or more autonomous aircraft which are connected to a ground station via one or more tethers. While various designs are investigated, two major crosswind concepts are currently considered by industry: the ground-generation also referred to as pumping-mode, and on-board-generation also referred to as drag-mode. On-board-generation AWES carry additional weight with the on-board
- 20 generator and propeller mass, as well as the heavier, conductive tether. This study focuses on the cyclic two-phase, ground-generation concept, as it is currently the main concept pursued by industry.

Ground-generation AWES generate power during the reel-out phases while the wing generates large lift forces and pulls the tether from a drum. <u>Various companies propose different reel-out pattern trajectories such as figure of eight or circular spirals</u>, which is investigated in this research. During the following reel-in phases a fraction of the previously generated energy is

25 consumed to return the aircraft back to its initial position and restart the cycle (Luchsinger, 2013). <u>The upward and downward motion during the production phases are called pumping cycles</u>. As a result, The power generated by such systems is inherently oscillating which could be offset using multiple devices in a wind farm setup or buffering the energy before feeding it into the grid (Malz et al., 2018; Faggiani and Schmehl, 2018).

Over the last years, two main AWES applications emerged. The first makes use of the mobile nature of the technology which

- 30 allows the deployment in inaccessible or remote places such as temporary mines or remote off-grid communities as these locations often rely on expensive diesel generators (SkySails Group GmbH, 2001; Kitepower B.V., 2016). added reference The second is the grid-scale integration of AWES, which requires upscaling the systems to compete with fossil and established renewable energy sources in the energy market. One example is Ampyx Power (Ampyx, 2020) which aims to re-power decommissioned offshore wind farms or deploy floating platforms (offshorewind.biz, 2018), expecting higher energy yield due
- 35 to better wind conditions, which in combination with <u>advantageous</u> design choices lead to lower levelized cost of electricity. Additionally, setting up AWES offshore allows for safer operation and is likely to be socially more accepted (Ellis and Ferraro, 2016).

Determining realistic performance of AWES is <u>difficult challenging</u> as <u>AWES</u> the flight trajectory depends on many variables that are not represented in simple models. Wind velocity profiles, aerodynamic coefficients, tether drag, aircraft mass and <u>AWES</u> size <u>impact</u> affect the flight <u>AWES</u> trajectory and therefore also the generated power. Using an optimization algorithm,

it is possible to account for implement these variables and determine optimal AWES performance.

We therefore investigate the scalability and design space of small to large-scale AWES, both offshore and onshore. Depending on the aircraft wing surface area, aerodynamic coefficients and the tether diameter, rated power ranges from $\overline{P}_{rated} = 145$ kW to 199 kW for $A_{wing} = 10 \text{ m}^2$ and $\overline{P}_{rated} = 2000 \text{ kW}$ to 3400 kW for $A_{wing} = 150 \text{ m}^2$. We compare the optimal system

45 performance subject to various wing mass of different aircraft masses for representative onshore and offshore wind conditions.

In comparison to the commonly used logarithmic wind speed profile, this the WRF-derived set of wind data includes the wind direction rotation with height and the complex range of profile shapes emerging from atmospheric stability. This includes almost constant wind velocity profiles associated with unsteady stratification, high sheer wind velocity profiles resulting from stable conditions, as well as non-monotonic wind velocity profiles including low level jets (LLJs). The power output of an

- 50 AWES not only depends on the wing size, but also the prevalent wind velocity profile shape and magnitude which result in distinct trajectories and operating altitudes. Therefore, a representative wind data set up to mid-altitudes, here defined as heights above 100 m and below 1500 m, is necessary to determine realistic AWES performance. This study relies on mesoscale numerical weather prediction models such as the Weather Research and Forecasting (WRF) model, which is well known for conventional WT siting applications (Salvação and Guedes Soares, 2018; Dörenkämper et al., 2020), as measuring wind
- 55 conditions at mid-altitudes is difficult due to reduced data availability aloft (Sommerfeld et al., 2019a). In comparison to the commonly used logarithmic wind speed profile, the WRF-derived set of wind data includes the wind direction rotation with height and the complex range of profile shapes emerging from atmospheric stability. This includes almost constant wind velocity profiles associated with unsteady stratification, high sheer wind velocity profiles resulting from stable conditions, as well as non-monotonic wind velocity profiles including low level jets (LLJs). To reduce the computational cost, 10-minute
- 60 average wind speed profiles were are clustered using the k-means clustering method described in Sommerfeld et al. (2020). We compare AWES performance for an onshore location in northern Germany near Pritzwalk (Sommerfeld et al., 2019b) and an offshore location at the FINO3 research platform in the North Sea. These wind clustered wind conditions were are implemented into the awebox (De Schutter et al., 2020) optimization framework which computes periodic flight trajectories that maximize average mechanical power output.
- In comparison to our previous studies (Sommerfeld et al., 2020), which derived onshore and offshore AWES power curves, this paper explores the AWES design space from small to to utility-scale. We aim at setting up-scaling design and mass targets, instead of a detailed system design. While other studies rely on simplified logarithmic wind speed profiles (De Schutter et al., 2019), high resolution L large eddy simulation (LES) (Haas et al., 2019) or reanalysis data sets (Schelbergen et al., 2020) to investigate general behavior, performance, trajectory or wake effects, we optimize AWES trajectory subject to realistic 10 minute mesoscale wind data, which allows better optimal performance prediction.
- The main contribution is the presentation of aerodynamic, mass and size scaling effects on representative ground-generation station AWES subject to realistic wind conditions and operating constraints. The here described results inform allow an informed decision-making regarding location-specific design, power estimation and scaling limitations.

The structure of the paper is as follows. Section 2 summarizes the onshore and offshore wind resource as well as the clustering results. For a detailed description of the WRF model and clustering algorithm see Sommerfeld et al. (2020). Section 3 briefly introduces the AWES model and optimization method as well as the implemented constraints and initialization. Section 4 compares the results for six AWES sizes with three different mass scaling assumptions and two sets of non-linear aerodynamic coefficients. We present, inter alia, <u>flight</u> trajectories, power curves and annual energy production estimates for a representative onshore and offshore location. Finally, Section 5 concludes the article with an outlook and motivation for future

80 work to continue to advance AWES towards commercial reality.

2 Wind data conditions

This study considers representative 10 min onshore (northern Germany, lat: $53^{\circ}10'47.00''$ N, lon: $12^{\circ}11'20.98''$ E) and offshore wind data (FINO3 research platform, lat: $55^{\circ}11,7'$ N, lon: $7^{\circ}9,5'$ E) wind data derived from 12 months of WRF simulations each. Both locations are highlighted by a black dot in Figure 1.

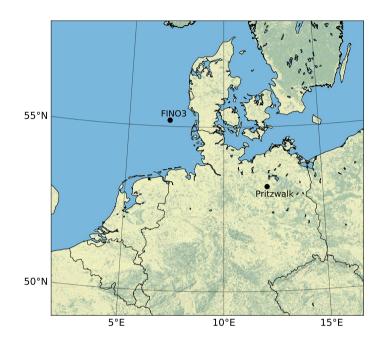


Figure 1. Topograph<u>yic</u> map of northern Germany with the representative onshore (Pritzwalk) and offshore (FINO3) locations highlighted with a black dots and labeled.

Both horizontal velocity components of the resulting mesoscale wind data set are clustered partitioned using a <u>k</u>-means clustering algorithm (Pedregosa et al., 2011). According to previous investigations (Sommerfeld et al., 2020), a small number of clusters with few representative profiles per cluster yield good power and AEP estimates at reasonable computational cost. Therefore, the wind velocity profiles were are grouped into k = 10 clusters from which the 5th, 50th and 95th percentile (sorted by <u>the</u> wind speed at 200 m) were are implemented into the optimization algorithm as design points to cover the entire annual wind regime.

The resulting average wind velocity profiles for each of the ten clusters, also known as centroids, are shown in the top row of Figure 2. For presentation purposes, only each centroid's wind speed magnitude, colored according to average wind speed up to 500 m, is shown. The complete set of clustered profiles profiles are shown in grey. The cluster average wind profile shapes

show wind shears typically associated with unstable and stable atmospheric conditions. They follow expected location-specific

- 95 trends with lower wind shear and higher wind speeds offshore (right) in comparison to onshore (left). The associated, colorcoded annual centroid frequency is shown in the <u>center middle row of Figure 2</u>. The <u>bottom subfigures summarize diagrams</u> in the bottom row illustrate the wind speed probability distribution at a reference height of $100 \le z \le 400$ m. We chose this reference height as a proxy for the wind speed at operating altitude, because an a priori estimation is impossible, and onshore and offshore power curves are almost identical using this reference wind speed. For a detailed description of the WRF model
- 100 and setup, the clustering process as well as the correlation between clusters and stability conditions see Sommerfeld et al. (2020). <u>Recent consensus among the AWES community defined the reference height as the pattern trajectory height, which is</u> the expected or actual time-averaged height during the reel-out (power production) phase (Airborne Wind Europe, 2021).

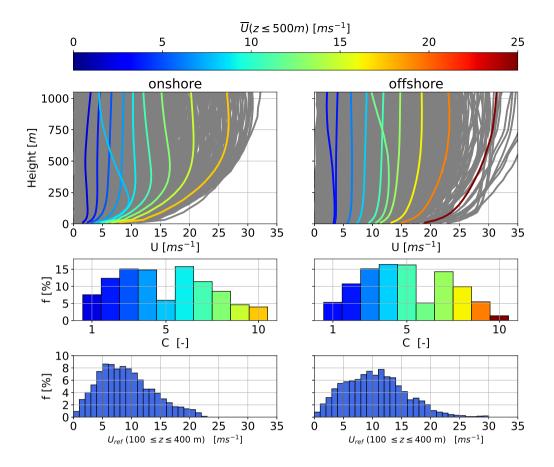


Figure 2. k-means clustered (k = 10) onshore (left column) and offshore (right column) annual cluster average wind speed profiles centroids (top row)-resulting from the k-means clustering process for k = 10 (top row). Underlying The range of WRF-simulated wind speed profiles is depicted in grey. The centroids are sorted, labeled and colored in according to average wind speed up to 500 m. The corresponding cluster frequency f for each cluster C is shown in the <u>center middle row</u>. The <u>bottom figures diagrams in the bottom row</u> show the wind speed probability distribution at a reference height of $100 \le z \le 400$ m. <u>Markus: left as is because I believe the figure needs to be</u> understandable without reading the entire text

3 AWES Trajectory optimization model

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Investigating the AWES scaling potential not only requires understanding of wind conditions at higher altitudes, but also of AWES power production, which is intrinsically linked to the aircraft's flight dynamics, as the AWES never reaches a steady state over the course of a power cycle. Forces and moments continuously change due to transition between reeling in and out within each pumping cycle as well as the changes in flight direction inherent to typical flight patterns, such as figures of eight or circular spirals, during the production phase. Additionally, constantly changing wind conditions over a wast height range require the aircraft to adapt its trajectory. Hence, power output estimation based on steady-state simplifications only 110 give a rough estimate, but can not describe the variation of system parameters or operating trajectory which determine power production, particularly for realistic, non-monotonic wind profiles. Therefore we make use of optimal control methods to compute power-optimal, dynamically feasible flight trajectories that satisfy operation constraints such as flight envelope and structural system limits. We compare the optimization results to a simplified quasi steady-state (QSS) engineering AWES model (QSM) similar to van der Vlugt et al. (2019) and Schmehl et al. (2013) (Sub-section 3.2) to verify our results and to highlight the difference between both models.

3.1 Model overview

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We compute ground-generation AWES power pumping cycles by solving a periodic optimal control problem which maximizes the cycle-average AWES power output \overline{P} . In periodic optimal control, the system state at the initial and final time of the trajectory must be equal, but are chosen freely by the optimizer. This methodology, implemented in the open-source software framework awebox (De Schutter et al., 2020), is used to generate power-optimal trajectories for single-wing ground-

generation AWES sizes with various variable wing area, mass and aerodynamic performance. Table 1 summarizes the imposed constraints for each system design.

The AWES model considers a 6 degree of freedom (DOF) rigid fixed-wing aircraft model with pre-computed quadratic lift, to account for stall effects, drag and pitch moment coefficients, which is controlled via aileron, elevator and rudder deflection

125 rates. For this scaling study, the Ampyx Power AP2 reference model (Ampyx, 2020; Licitra, 2018; Malz et al., 2019) added reference serves as a base from which the aircraft size and mass as well as aerodynamic coefficients were are scaled (see Sections 3.4 and 3.6).

While the ground station dynamics are not explicitly modeled, constraints on tether <u>reeling</u> speed, acceleration and jerk are implemented to ensure a realistic operating envelope. For this study a reel-out to reel-in <u>speed</u> ratio of $\frac{2}{3}$ ($\frac{v_{out}}{v_{in}} = \frac{10 \text{ ms}^{-1}}{15 \text{ ms}^{-1}}$ 130 $\frac{\dot{l}_{out}}{l_{in}} = \frac{10 \text{ ms}^{-1}}{15 \text{ ms}^{-1}}$) was chosen. Tether acceleration $\ddot{l}_{max} = 10 \text{ ms}^{-2}$ and tether jerk $\ddot{l}_{max} = 100 \text{ ms}^{-3}$ were are limited to simulate generator torque constraints. Tether tension is defined by its diameter and maximum strain stress. The tether diameter <u>d</u> has been chosen such that maximum average cycle power is achieved at an approximate wind speed of 10 ms^{-1}.

For a more detailed description of the model and the optimization algorithm see Sommerfeld et al. (2020); Leuthold et al. (2018); De Schutter et al. (2019); Bronnenmeyer (2018); Horn et al. (2013); Haas et al. (2019).

135 3.2 Quasi-steady state reference model

To contextualize the optimization results, a quasi-steady state model (QSS QSM) based on Loyd's ideal crosswind approximation operation (Loyd, 1980) is introduced. This model has been adapted generalized by Schmehl et al. (2013) to include misalignment losses arising from misalignment of the tether and wind velocity vector. The aircraft position is described in the spherical coordinates by the distance from the ground station, the elevation angle $\underline{\theta}_{\underline{\varepsilon}}$ and azimuth angle relative to the wind velocity

140 vector. It neglects aircraft and tether mass and assumes a quasi-steady flight state, with the wing moving directly cross-wind

with zero azimuth angle $\phi = 0$ relative to the wind direction. Dividing the tether reeling speed $\underline{v_t}\underline{\dot{l}}$ by the wind speed is non-dimensionalized as defines the reeling factor

$$f = \underline{v_t} \underline{\dot{l}} / U. \tag{1}$$

with an optimal value of $f_{opt} = 1/3 \cos \theta \varepsilon \cos \phi$ (Argatov et al., 2009). <u>added ref</u> Equation (2) estimates <u>optimal maximum</u> 145 power P_{optmax} as a function of wind speed U at altitude z and the resultant aerodynamic force coefficient c_R (see Equation (3)) which is calculated from the aerodynamic lift σ_{c} and total drag coefficient σ_{c} of all airborne components

(3)), which is calculated from the aerodynamic lift $c_{\rm L}$ and total drag coefficient $c_{\rm D,total}$ of all airborne components.

$$P_{\underline{\text{opt}}\underline{\text{max}}} = \frac{\rho_{\text{air}}(z)}{2} U(z)^3 c_{\text{R}} \left(\frac{c_{\text{R}}}{c_{\text{D,total}}}\right)^2 f_{\text{opt}} \left(\cos\underline{\theta} \underline{\varepsilon}\cos\phi - f_{\text{opt}}\right)^2$$
(2)

$$c_{\rm R} = \sqrt{c_{\rm L}^2 + c_{\rm D,total}^2} \tag{3}$$

Tether drag is included in the total AWES drag coefficient $c_{D,total}$ according to Equation (5) a simplified estimation (Houska and Diehl, 2007; Argatov and Silvennoinen, 2013; van der Vlugt et al., 2019; Schmehl et al., 2013).

Maximizing Increasing the power output P_{optmax} is mainly a function of can be achieved by improving $c_{\text{R}}^3/c_{\text{D,total}}^2$ and wind speed U at height z as well as tether length l_{tether} , which determine the elevation angle $\underline{\theta}_{\varepsilon} = \arcsin(\frac{z}{l_{\text{tether}}})$ and tether assosciated losses. A linear approximation of the standard atmosphere yields air density $\rho_{\text{air}}(z)$ at altitude z (Champion et al., 1985)

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$$\rho_{\rm air}(z) = 1.225 \, {\rm kgm}^{-3} - 0.00011 \, {\rm kgm}^{-4} z.$$
 (4)

The total drag coefficient $c_{D,total}$ determines represents the air resistance aerodynamic drag of the entire AWES in crosswind motion. It depends on the tether diameter d_{tether} and length l_{tether} , as well as the wing area A_{wing} . The aerodynamic drag coefficient of the wing $c_{D,wing}$ is defined by the wing shape depends on the angle of attack and the shape of the wing. We consider a cylindrical tether with constant diameter and an aerodynamic tether drag coefficient $c_{D,tether} = 1.0$. The tether drag coefficient could even be higher for braided tethers. For the sake of simplicity, tether slope inclination with respect to the wind direction is not considered in the drag calculation, which leads to an over estimation of tether drag. A more accurate tether model would further include the wind speed variation with height. Assuming a uniform wind field, the line integral along the tether results in a total effective drag coefficient of (Houska and Diehl, 2007; Argatov and Silvennoinen, 2013; van der Vlugt et al., 2019):

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$$c_{\mathrm{D,total}} = c_{\mathrm{D,wing}} + \frac{1}{4} \frac{d_{\mathrm{tether}} l_{\mathrm{tether}}}{A_{\mathrm{wing}}} c_{\mathrm{D,tether}}$$
 (5)

Both the QSS QSM and the optimization model are subject to the same constraints (Table 1). The optimal power of the QSS QSM model is estimated by varying tether length up to 2000 m for every given wind profile (Section 3.3) and applying the above described tether drag and elevation losses. The same minimal operating altitude as for the optimization model is enforced. The QSS QSM-predicted power used for reference in Sub-section 4.3 are is the highest power for a given wind

170 profile. Therefore, optimial operating height is the height at which the highest power is calculated, see previous publication (Sommerfeld et al., 2020).

3.3 Wind boundary condition profile implementation

The 2D horizontal wind velocity profiles were are clustered partitioned into k = 10 clusters. Three representative profiles from each cluster as well as each cluster's centroid, rotated such that the main wind direction \boldsymbol{w} points in positive x-direction and the

- 175 deviation v from it transverse velocity component points in positive y-direction, wereare implemented into the optimization algorithm as boundary conditions. This assumes that the investigated AWES can operate independent of wind direction and are not restricted to a certain direction. This way the main wind direction of every profile points in the same direction, simplifying the comparison between different wind velocity profiles. omnidirectional AWES operation, which simplifies the comparison of results. We interpolate the x-direction component u and y-direction component v components using Lagrange polynomials to
- 180 obtain a twice continuously differentiable function representation of the wind velocity profiles, which is necessary to formulate an optimal control problem that can be solved with the gradient-based nonlinear programming (NLP) solver IPOPT (Waechter and Laird, 2016).

3.4 Aircraft scaling

Aircraft mass m and inertia J are scaled relative to the Ampyx AP2 reference model (Licitra, 2018; Malz et al., 2019; Ampyx, 2020) according to simplified geometric scaling laws relative to wing span b_{scaled} (see Equations (6) and (7)).

$$m_{\rm scaled} = m_{\rm ref} \left(\frac{b_{\rm scaled}}{b_{\rm ref}}\right)^{\kappa} \tag{6}$$

$$\mathbf{J}_{\text{scaled}} = \mathbf{J}_{\text{ref}} \left(\frac{b_{\text{scaled}}}{b_{\text{ref}}}\right)^{\kappa+2} \tag{7}$$

We investigate the impact of positive and negative scaling effects by varying the mass scaling exponents κ ranges from between 2.7 toand 3.3. An exponent of 3 represents pure geometric scaling (North et al., 2007) according to the square-cube
190 law, while κ = 2.7 implies positive scaling effects and weight savings with size, while κ = 3.3 assumes negative scaling. A review of available literature shows that anticipated AWES scaling exponents vary between κ = 2.2 - 2.6 (grey area), shown in Figure 4. We assume more conservative mass scaling for the purpose of this investigation.

Makani's openly published technical reports describe their "M600 SN6" as well as their MX2 (*Oktoberkite*) design, which redesigned is a redesign of the M600 airframe to overcome some of its shortcomings and produce $\overline{P}_{MX2} = 600$ kW at a wind

speed of $U_{MX2-ref} = 11 ms^{-1}$ at operating height (Echeverri et al., 2020). Note that Makani's on-board-generation concept 195 is inherently heavier than the ground-generation concept considered here, because of propellers, generators and supporting structures onboard the aircraft. The intended original M600 design specified was designed for a mass of 919 kg., which corresponds to an AP2 mass scaling exponent of $\kappa = 2.72$. The as-built M600 had a wing area of $A_{\text{wing}} = 32.9 \text{m}^2$ and a mass of $m_{\rm M600} = 1730.8$ kg which is almost double the design value. If we scale the AP2 reference aircraft to the same wing area and mass, the corresponding mass scaling exponent is $\kappa = 3.23$. The airframe of the improved MX2 design aimed at 200 $m_{\rm MX2} = 1852 \rm kg$ for a wing area of $A_{\rm MX2} = 54 \rm m^2$, equivalent to $\kappa = 2.719$ relative to the AP2 reference. Similarly, wind turbine (WT) mass scales with an exponent slightly below 3 based on rotor diameter (Fingersh et al., 2006).

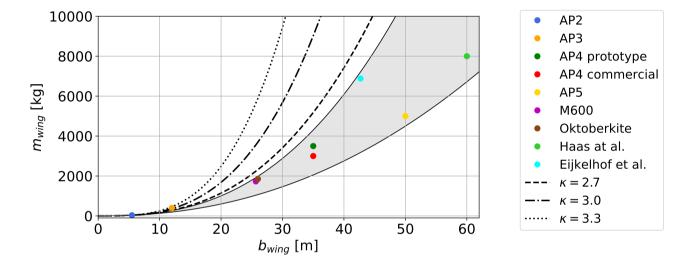


Figure 3. Curve fit of Published sizing study AWES aircraft masses (Haas et al., 2019; Kruijff and Ruiterkamp, 2018; Eijkelhof et al., 2020; Ampyx, 2020; Echeverri et al., 2020). For these data mass scales within a scaling exponent range of $\kappa = 2.2 - 2.6$ (grey area). The investigated, more conservative mass scaling exponents between $\kappa = 2.7 - 3.3$, are depicted by dashed, dash-dotted and dotted lines.

3.5 **Tether model**

The tether is modeled as a straight, cylindrical solid rod with constant diameter which can not support compressive forces. 205 This is a good reasonable assumption when tether tension is high during the power production phase of the power cycle. Total tether drag is proportional to tether diameter d_{tether} , which scales with tether tension and therefore wing area, assuming constant tensile strength. and tether length ltether. Both scale with tether tension, assuming a constant tensile strength, and therefore aircraft size and wind speed (see Subsection 4.5). The tether drag is approximated by dividing the tether into multiple elements segments (here $n_{\text{tether}} = 10$) and calculating the apparent wind speed velocity at each element individually.

assuming We assume a constant tether drag coefficient of $C_{\rm D}^{\rm tether} = 1$, which is the drag coefficient of a smooth cylindrical 210 object at higher Reynolds numbers (typical for AWE applications) (Hoerner, 1965) and could even be higher for braided

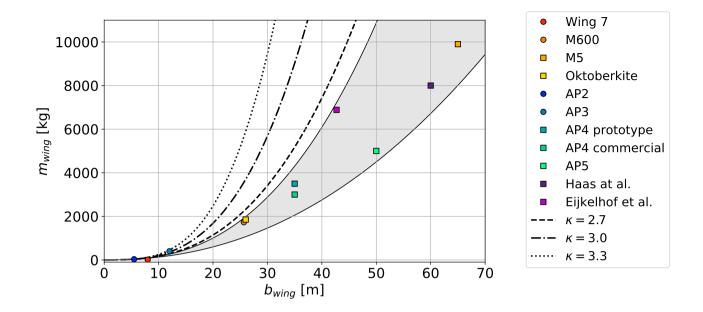


Figure 4. <u>Published AWES aircraft masses</u> (Haas et al., 2019; Kruijff and Ruiterkamp, 2018; Eijkelhof et al., 2020; Ampyx, 2020; Echeverri et al., 2020). For these data, the scaling exponent ranges between $\kappa = 2.2 - 2.6$ (grey area). The here investigated, more conservative mass scaling exponents between $\kappa = 2.7 - 3.3$, are depicted by dashed, dash-dotted and dotted lines. **Roland: New work by you or published** work? Markus: This work

tethers. Each elements tether drag The drag contribution of each segment is then equally divided between the two its endpoints and finally transferred propagated to either the aircraft or ground station. Please refer to (Leuthold et al., 2018) for more details. This leads to an underestimation of total tether drag at the aircraft (Leuthold et al., 2018). Similarly, the total tether weight W_{tether} , calculated with a constant material density of $\rho_{\text{tether}} = 970 \text{ kgm}^{-3}$, is distributed evenly between the aircraft and ground station. Tether force constraints are chosen such that the system's rated power is achieved for a logarithmic wind speed profile with $U_{\text{sizing}}(z = 200 \text{ m}) \approx 10 \text{ms}^{-1}$ at $U_{\text{sizing}}(100 \le z \le 400 \text{ m}) \approx 10 \text{ms}^{-1}$, assuming a logarithmic wind speed profile , similar to wind at hub height for conventional wind turbines. Therefore, the tether diameter of every AWES design (Table 1) is derived from the maximum tether stress $\sigma_{\text{tether}} = 3.6 \cdot 10^9 \text{ Pa}$ and a safety factor $SF_{\text{tether}} = 3$.

The Gground station is not explicitly modeled, instead hypothetical tether <u>reeling</u> speed and acceleration constraints are imposed, mimicking motor torque and rotational speed and motor torque limitations. Maximum reel-out speed is limited to $v_{out}15\dot{l}_{in} = 10 \text{ ms}^{-1}$ and reel-in speed to $v_{out} = 10\dot{l}_{in} = 15 \text{ ms}^{-1}$, resulting in a reel-out to reel-in ratio of $\frac{2}{3}$ which is assumed to be within design limitations of the winch. This limits the mechanical, instantaneous power that each ground-generation AWES can generate $P_{\text{inst.}}^{max} = F_{\text{tether}}^{max} v_{out}\dot{l}_{out}$. A maximum tether acceleration of $\ddot{l} = 10 \text{ ms}^{-2}$ is imposed to comply with generator torque limits.

3.6 Aerodynamic scaling

Figure 5 compares the aerodynamic performance of the AP2 wing with and without a 500 m tether to a high-lift wing. The solid lines show the aerodynamic coefficients of the unterthered aircraft (l = 0m) and the dashed lines the ones of the tethered aircraft with a tether length of l = 500 m. Lift $C_{\rm L}$ (a), drag $C_{\rm D}$ (b) and pitch moment $C_{\rm m}$ coefficients (c) and glide ratio are 230 depicted as a functions of angle of attack (e). Lift-over-drag is shown in (d). The bottom right figure Diagram (f) displays the $\frac{C_{\rm R}^3}{C_{\rm T}^2}$ ratio (Equation (2)) which determines the theoretical maximum power of any crosswind AWES as defined by Equation (2) (Loyd, 1980; Diehl, 2013). Makani's reports Echeverri et al. (2020) mention that two shortcomings of their the original M600 design were the overestimation of $C_{\rm L}^{\rm max}$ and underestimation of $C_{\rm D}$ Roland: Including tether drag? Markus: not specified, but I assume they refere to operation, so yes with tether drag, further justifying this comparison and prompting a more 235 conservative estimation of practical aerodynamic coefficients. The aerodynamic coefficients of the AP2 reference model were identified by Licitra (2018); Malz et al. (2019) in AVL (Drela and Youngren, 2016) and confirmed through CFD analyses by Ampyx Power (Vimalakanthan et al., 2018) and during untethered test flights. Modifications to the AP2 aerodynamic reference model were are implemented to assess the impact of improved aerodynamics on AWES performance (labeled HL for high-lift). This is achieved by shifting the $C_{\rm L}$, $C_{\rm D}$ and $C_{\rm m}$ as if high-lift devices, such as fixed trailing-edge flaps and fixed leading-edge slots, were attached (Kermode et al., 2006; Lee and Su, 2011; Hurt, 1965; Scholz, 2016). This is achieved by increasing the 240 lift, drag and moment and coefficients at $\alpha = 0$ and increasing the stall angle. The high-lift configuration does not represent a

specific design, but an arbitrary improvement in aerodynamic efficiency, which is here defined as higher lift-to-drag ratio, in comparison to the reference AP2 data. Lift and drag at zero angle of attack are increased, stall is delayed, and pitch moment decreased. While both airfoils have comparable optimal glide ratios, the Loyd's optimal power ratio power harvesting factor at zero elevation an azimuth angle (Loyd, 1980)

$$\zeta = \frac{P}{P_{\text{area}}} = \frac{\overline{P}}{\frac{1}{2}\rho_{\text{air}}A_{\text{wing}}}U(z)^3 \le \frac{4}{27}c_{\text{R}}\left(\frac{c_{\text{R}}}{c_{\text{D}}}\right)^2 \tag{8}$$

is almost twice as high for the high-lift airfoil. The power harvesting expresses the estimated AWES power *P* relative to the total wind power through an area the same size as the wing P_{area} . It can be derived from (2) by setting the elevation angle ε and the azimuth angle ϕ to zero. An extreme value analysis results in an optimal reel out speed \dot{l} of 1/3 of the wind speed 250 $\frac{Y(1)}{2}$ and $\zeta_{max} = \frac{4}{27} c_R \left(\frac{c_R}{e_D}\right)^2$. U(z) is the wind speed and $\rho_{air}(z)$ the air density at operating altitude. Stall effects were are implemented for both the AP2 reference model (blue) as well as the high-lift (HL - orange) model by formulating fitting the lift curve to a quadratic lift coefficient function (see Figure 5). As a result, the lift coefficients deviate slightly in the linear lift region at lower angle of attack.

3.7 Constraints

255 As previously mentioned, the AWES model solves a constraint constrained optimal control problem to maximize average cycle-power of a single 6 DOF tethered aircraft connected to the ground station via a single rigid straight inelastic tether.

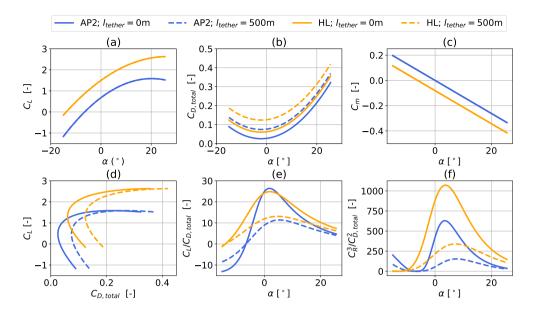


Figure 5. The dashed lines in (b) show $C_{D,total}$ (aircraft + 500 m tether) while the solid lines show $C_{D,wing}$ (only aircraft). Aerodynamic lift C_L (a) , drag C_D (b), and pitch moment C_m coefficients (c), each and with (dashed line) and without tether drag (solid line) as a function of angle of attack for AP2 (blue) (Licitra, 2018; Malz et al., 2019) and high-lift (HL) (orange) configuration. The bottom sub-figures diagrams display Lift-over-drag(d) lift as a function of total drag, lift-to-drag ratio over angle of attack (e) and $c_R^3/c_{D,total}^2$ over angle of attack according to Loyd (Loyd, 1980). HL coefficients are derived by modifying the AP2 reference model as if arbitrary high-lift devices ; e.g. flaps and slats were attached.

Each run optimizes the trajectory during the production cycle pumping cycle of an AWES at a fixed size subject to varying wind conditions for a given wind field (wind velocity as a function of altitude). The wind field is assumed to be constant for every optimization. Constraints include include system dynamics, material properties, aircraft (Sub-section 3.6) and ground station hardware constraints as well as flight envelope limitations listed in Table 1. These limitations include minimum and maximum operating heights ($z_{operation}^{min}$ and $z_{operation}^{max}$), maximum acceleration a_{flight}^{max} measured as multiples of g a_{flight}^{max} as well as a maximum tether length l_{tether}^{max} to maintain safe operation. More information on the model and constraints can be found under in De Schutter et al. (2020) and the therein referenced publications. The number of loops n_{loop} within the reel-out phase

260

of every each pumping cycle is fixed to 5.

265 The maximum tether stress and force, from which the tether diameter is calculated, together with the periodicity constraint are some of the most important path constraints. Ground station hardware limitations such as torque and acceleration dynamics are not explicitly modeled, but implemented as tether reeling speed and acceleration constraints. A Fixed angle of attack α and side slip angle β range ensures operation within realistic bounds. Angle of attack $-30 \le \alpha \le 30$ and side slip angle $-15 \le \beta \le 15$ of the wing constraints ensure operation within realistic bounds. However, neither angular constraint is active

270 during flight, because the optimizer tries to achieve an angle of attack close to the ideal harvesting factor ζ maximum of $C_{\rm L}$,

Table 1. List of investigated AWES design parameters and selected, important system constraints for the six investigated designs with different wing sizes $(10m^2 \le A \le 150m^2)$ with the original AP2 design as reference. The two different aerodynamic configurations (AP2 and HL) determine tether diameter *d* and maximum tether force $F_{\text{tether}}^{\text{max}}$. HL and AP2reference aerodynamic coefficients

Parameter		AP2	design 1	design 2	design 3	design 4	design 5	design 6			
	$A_{\underline{\text{wing}}???}$ [m ²]	3	10	20	50	80	100	150			
Aircraft	$c_{\rm wing}$ [m]	0.55	1.00	1.41	2.24	2.83	3.16	3.87			
	$b_{\rm wing}$ [m]	5.5	10	14.1	22.4	28.3	31.6	38.7			
	AR [-]	10	10	10	10	10	10	10			
	$m_{ m kite}(\kappa=2.7)$ [kg]	36.8	185	471	1,624	3,062	4,139	7,155			
	$m_{\rm kite}(\kappa=3.0)$ [kg]	36.8	221	626	2,473	5,005	6,995	12,850			
	$m_{\rm kite}(\kappa=3.3)$ [kg]	36.8	265	830	3,767	8,180	11,821	23,079			
	α [°]	[-10:30]									
	β [°]	[-15 : 15]									
	$l_{\underline{\text{tether}}}^{\max}$ [m]	2000									
Tether	$\dot{l}_{\underline{\text{tether}}} [\text{ms}^{-2}]$	[-15 : 10]									
	$\ddot{l}_{\underline{\text{tether}}} [\text{ms}^{-2}]$	[-15 : 10]									
	$\ddot{l}_{\underline{tether}}^{\max} [ms^{-3}]$	[-20 : 20]									
	$ ho^{ m tether} [m kgm^{-3}]$	970									
	$\sigma_{\rm max}^{\rm tether}$ [Pa]	3.6 ± 10^9									
	SF^{sigma} [-]	3									
	$d_{\underline{tether}}(AP2) \ [mm]$		5.5	7.8	12.3	15.5	20	21.7			
	$d_{\rm \underline{tether}}({ m HL}) \ [{ m mm}]$		7.2	10.2	16.1	20.6	23	28.3			
	$F_{\text{tether}}^{\text{max}}(\text{AP2})$ [kN]		34	60	136	241	377	456			
	$F_{\rm tether}^{\rm max}({ m HL})$ [kN]		46	94	241	416	499	738			
flight envelope	$z_{ m operating}^{ m min}$ [m]		55	60	75	90	100	125			
	$z_{ m operating}^{ m max}$ [m]	1000									
	$v_{\rm flight}^{\rm max} [{\rm ms}^{-1}]$	80									
	$a^{\max} \left[\underline{g} \underline{\mathrm{ms}}^{-2}\right]$	12 <u>× g</u>									
	$n_{ m loop}$ [-]	5									
Initialization	$\underline{N}\underline{n}_{ m loops}$	5									
	ε [°]	30									
	$l^{\mathrm{init}}_{}{\underline{tether}}}\left[\mathrm{m} ight]$		500	535	643	750	821	1000			

 $C_{\rm D}$ and $C_{\rm m}$ (see Figure 5). Due to weight and drag effects, the actual angle of attack is closer to $\alpha \approx 10^{\circ}$ during reel-out for the majority of wind speeds.

3.8 Initialization

The AWES dynamics are highly non-linear and therefore result in a non-convex optimal control problem which possibly has

- 275 multiple local optima. Therefore, the particular results generated by a numerical optimization solver can only guarantee local optimality, and usually depend on the chosen initialization The optimization is initialized with a circular trajectory based on a fixed number of $n_{\text{loop}} = 5$ loops at a 30° elevation angle in positive *x*-direction and an estimated aircraft speed of $v_{\text{init}} = 10 \text{ ms}^{-1}$. Previous analyses showed that the convergence of large AWES highly strongly depends on initial tether length. Larger systems become are less sensitive to tether drag, because lift_to_tether drag ratio scales linearly with wing span.and
- 280 hence can drag along a longer tether. Larger and heavier aircraft have a higher moment of inertia and hence have a larger turning radius, requiring a longer tether. Therefore, Initial tether length l_{tether}^{init} is increased linearly with aircraft wing area (see Table 1) to improve the optimization process.

In order to solve the highly nonlinear optimization problem, an appropriate initial guess is generated using a homotopy method similar to those detailed in Gros et al. (2013); Malz et al. (2020). This technique gradually relaxes the problem from simple tracking of circular loops to the original nonlinear path optimization problem where the previous result serves as an initial guess for the following problem. An initial circular path, which is determined from the tether length guess and estimated flight speed, is transformed into a periodic helix-like trajectory. Several initial tether lengths were investigated to determine a feasible initial path depending on system mass, system size and wind speed. Initial tether lengths needs to increase with system size and wind speed. The resulting problem is formulated in the symbolic modeling framework CasADi for Python (Andersson et al., 2019, 2012) and solved using the NLP solver IPOPT (Wächter and Biegler, 2006) in combination with the linear solver

et al., 2019, 2012) and solved using the NLP solver IPOPT (Wächter and Biegler, 2006) in combination with the linear solver MA57 (HSL, 2020).

4 Results and discussion

We compare 6 six AWES sizes with three different mass properties and two sets of nonlinear aerodynamic coefficients each to investigate the AWES design space and upscaling potential. Furthermore, we contrast AWES the performance at representative onshore (Pritzwalk in northern Germany) and offshore locations (FINO3 research platform in the North Sea) based on one year of WRF simulated and k-means clustered wind data. To that end, we show representative optimized trajectories (Subsection 4.1) and compare typical operating altitudes and tether lengths (Subsection 4.2). Subsection 4.3 analyses AWES power curves for each design and determines an AWES power coefficient based on swept area and wing chord. From this we derive the annual energy production (AEP) in Subsection 4.4 for each location and system configuration. We examine the predicted power losses

300 (Subsection 4.6) due to tether drag. Finally, we establish an upper limit of the weight-to-lift ratio and compare tether drag forces in Subsection 4.5.

4.1 Flight trajectory and time series results

The trajectories in sub-figure b) and d) of Figure 6 (b) and 6 (d) depict the local optima of the highly non-linear model and optimization problem for AWESs designs 3 with a wing area of $A_{\text{wing}} = 50 \text{ m}^2$, both reference (AP2, solid lines) and HL

(dashed lines) aerodynamic coefficients and a scaling coefficient of $\kappa = 3$. The trajectories seem reasonable, are within the set

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constraints and are consistent with other studies (De Schutter et al., 2019; Sommerfeld et al., 2020) which use the same model. Sub-figure a) of Figure 6 (a) shows the vertical wind speed profiles U over altitude z with the operating region highlighted in color. Any deviation from the WRF data in grey is caused by the interpolation with Lagrange polynomials during the implementation process described in Subsection 3.3. The hodographs in sub-figure c) Figure 6 (c) show a top view of the

310 rotated <u>horizontal</u> wind velocity components u and v up to a height of 1000 m which follow the expected clockwise rotation with altitude (Stull, 1988). The average wind speed between 100 m $\leq z \leq 400$ m is used as reference wind speed U_{ref} , as it is a good enough proxy for conditions at operating height.

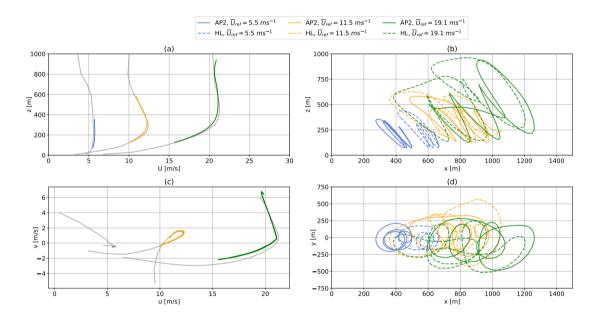


Figure 6. Optimization results of <u>one pumping cycle for the</u> ground-generation AWES with a wing area of $A_{wing} = 50 \text{ m}^2$, mass scaling exponent $\kappa = 3$ for both AP2 reference (solid lines) and high-lift HL (dashed lines) aerodynamic coefficients at various WRF-generated wind conditions. Sub-figure Diagrams (a) and (c) depict representative horizontal onshore wind speed profiles and their hodographs of wind velocity up to 1000 m. The deviation of the colored lines is caused by the implementation of discrete WRF-simulated data points using Lagrange polynomials. Sub-figure Diagrams (b) and (d) show the optimized trajectories in side and top view.

Trajectories at higher wind speeds and above rated power deviate noticeably from the expected trajectory, which manifest trajectories computed at lower wind speeds. The system optimization algorithm tries to de-power the aircraft by moving it out

315 of the power zone of the wind window , which is the low elevation angle zone directly downwind. As a result, the trajectory either shifts upwards, increasing $\theta_{\mathcal{E}}$, or sideways, increasing ϕ , as can be seen from Equation (2), or perpendicular to the main wind direction, to stay within the tether force, tether reeling speed and flight speed constraints, while still maximizing average cycle power. Subsection 4.2 further analyzes the trend towards longer tethers and higher operating altitude with increasing wind speed. , which can be seen here as distance from the origin.

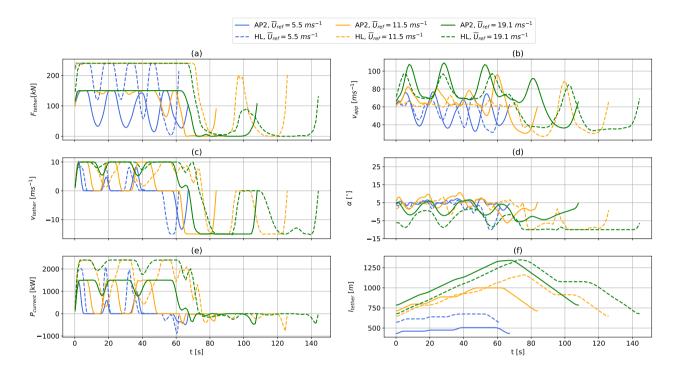


Figure 7. Time series of optimized ground-generation AWES with a wing area of $A_{\text{wing}} = 50 \text{ m}^2$, mass scaling exponent $\kappa = 3$ for both AP2 reference (solid lines) and high-lift HL (dashed lines) aerodynamic coefficients at various WRF-generated wind conditions. The corresponding trajectories are shown in Figure 6. The sub-figures show tether force F_{tether} (a), tether speed v_{tether} (c) and instantaneous power P_{current} (e), as well as apparent wind speed v_{app} (b), angle of attack α (d) and tether length l_{tether} (f).

- Figure 8 describes shows the time-dependent attitude temporal evolution of the previously analyzed trajectories during the operating cycle cycle trajectories depicted in Figures 6 (b) and 6 (d). Cycle time The cycle duration varies with wind speed and system configuration. A possible explanation is that At lower wind speeds (U_{ref} = 5.5ms⁻¹), the better aerodynamics of the HL configuration lead to higher flight speed and therefore shorter time to complete the cycle, as can be seen for lower wind speeds (U_{ref} = 5.5 ms⁻¹, blue). At higher wind speeds however, the system reduces its flight speed to stay within constraints, the HL configuration needs to reduce the flight speed to stay within constraints and also requires a much longer reel-in phase,
 - both leading to a longer cycle time. Because of the initialization with a simple circular trajectory and the fixed number of loop maneuvers, which are maintained during the optimization process, the reel-in period sometimes shows a loop maneuver as

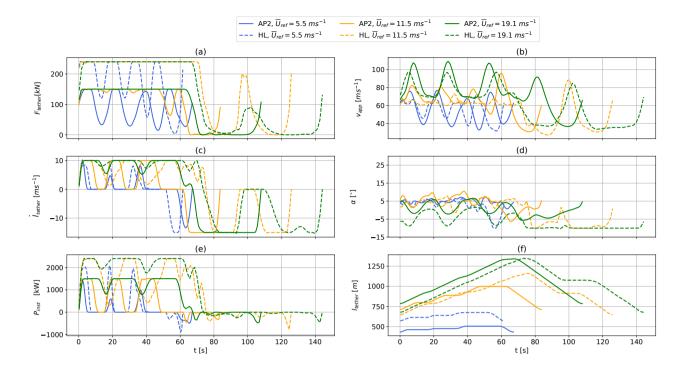


Figure 8. Time series of one optimized pumping cycle for the ground-generation AWES with a wing area of $\underline{A_{\text{wing}}} = 50 \text{ m}^2.$ mass scaling exponent $\kappa = 3$ for both AP2 reference (solid lines) and high-lift HL (dashed lines) aerodynamic coefficients at various WRF-generated wind conditions. The corresponding trajectories are shown in Figure 6. The sub-figures show tether force F_{tether} (a), tether speed v_{tether} (c) and instantaneous power $P_{\text{currentinst}}$ (e), as well as apparent wind speed v_{app} (b), a The diagrams show tether force F_{tether} (a), apparent wind speed v_{app} (b), tether reeling speed v_{tether} (c), angle of attack α (d) and instantaneous power R

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well. This can be seen for HL $U_{ref} = 11.5$ and $19.1 \text{ [ms}^{-1]}$ at around 100 seconds. The algorithm maintains a fixed number of loops maneuvers, resulting in unnecessary and probably less optimal trajectories during the reel-in period, as can be seen for HL $U_{ref} = 11.5$ and $19.1 \text{ [ms}^{-1]}$ at around 100 seconds. Our previous, unpublished analyses, utilizing the same model, showed that AWES power output seems to be fairly insensitive to both number of loops (here $n_{\text{loop}} = 5$ for all setups) and flight time. However, This needs to be verified and compared to other models and real experiments.

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The optimizer aims to achieve a constant, maximum tether force F_{tether} (Figure 8 (a)) during the reel-out period and vary tether reel-out speed (Figure 8 (c)) to maximize power (e) (Figure 8 (e)). This is achieved by varying the angle of attack (d) (Figure 8 (d)), while trying to stay close to the optimal $C_{\rm R}^3/C_{\rm D,total}$ (Figure 5 (f)). At high wind speeds the angle of attacked has to decrease to stay within tether tension constraints (orange and green lines). The diverse AWES trajectories generate are characterized by periodic evolutions of aerodynamic forces and tether tension. In the production phase, the tether reels-out and the aircraft follows an almost circular pattern, which leads to deceleration of the aircraft during the ascent and acceleration during the descent, due to gravity. To maintain tether tension, tether reeling speed decreases to zero. At lower wind speeds the

- aircraft can not produce sufficient lift force to pull the tether and overcome gravity during the ascent within each loop of the production cycle. As a result, tether force (Figure 8 (a)) decreases together with apparent wind speed v_{app} (Figure 8 (b)), tether reeling speed $v_{apptether}$ (b) (Figure 8 (c)) and instantaneous power $P_{currentinst}$ (see $U_{ref} = 5.5 \text{ [ms^{-1}]}$, blue) (Figure 8 (e)). Even at higher wind speeds (see $U_{ref} = 11.5 \text{ [ms^{-1}]}$, orange) the tether reeling speed drops to zero at during the ascent. As a consequence, the generated power also drops to zero and ramps up again (Figure 8 (e)), leading to grid feed-in challenges.
- Additionally, aerodynamic loads drop to almost zero during the recovery reel-in phase as the aircraft returns to its initial position and the tether is reeled in. To reduce the power losses and decrease the reel-in time, tether reeling speed quickly reaches its minimum of $v_{\text{tether}} = 15 \text{ms}^{-1}$. To alleviate this inherent intermittency Buffering the energy or coupling multiple, phase-shifted AWES in a wind farm setup would be beneficial (Faggiani and Schmehl, 2018; Malz et al., 2018), to alleviate this inherent power fluctuation between the production (reel-out) and the consumption (reel-in) phases.
- At lower wind speeds, aerodynamic forces oscillate about a constant base load during , the tether force (Figure 8, a blue line) approaches the maximum tether force during the peaks of the reel-out phase. With increasing wind speed, aerodynamic forces saturate due to tether tension constraints (Table 1), leading to increasing periods of constant, maximum tension. However, tether force troughs decrease even further with tether length, due to increased total system weight. Figure 9 gives an insight into the tether load cycles during the reel-out phases of an AWESs with a wing area of A_{wing} = 50 m². , both AP2 and HL aerodynamic coefficients and mass scaling of κ = 2.7, 3.0, 3.3. Average time between troughs (Figure 9 (a)) slightly increases with κ due to increased aircraft inertia, but remains almost constant with wind speed U_{ref}(100 m ≤ z ≤ 400 m). The relative reduction

of tether tension troughs (b) $\Delta F_{\text{tether}}/F_{\text{tether,max}} = |F_{\text{trough}} - F_{\text{tether,max}}|/F_{\text{tether,max}}$ decreases with wind speed as the apparent wind speed at the wing increases (Figure 9 (b)). As expected, this decrease is more pronounced for heavier AWES. Higher aerodynamic efficiency (HL circular marker and dotted line) increases performance and smooths out the troughs. The

360 lightest configuration achieves constant reel-out tension at around rated wind speed of $U_{ref} = 10ms^{-1}$, which is the lowest wind speed at which the AWES can produce its rated power P_{rated} , while the heaviest design requires higher wind speeds of about $15ms^{-1}$.

4.2 Tether length and operating altitude

One of the major value propositions of AWESs is that they can tap into wind resources beyond the reach of conventional wind turbines. However, The choice of optimal operating height highly strongly dependents on the vertical wind speed profile and system design. Two opposing effects influence the optimal operating height. On the one hand, an increase in altitude is generally associated with an increase in wind speed and therefore produced power. On the other hand, higher altitudes require a longer tether which results in higher drag losses or <u>an</u> increase <u>of</u> the elevation angle which increase "cosine" losses (Diehl, 2013) ,or both.

Figure 11 shows a trend towards longer average tether lengths \bar{l}_{tether} (top) and higher average operating altitudes $\bar{z}_{operating}$ (centermiddle) with increasing system size for a representative scaling exponent of $\kappa = 3$ (see Equations (6) and (7)) and wind speed. We chose U_{ref} to be the average wind speed between 100 m $\leq z \leq 400$ m as we previously found that this range is a

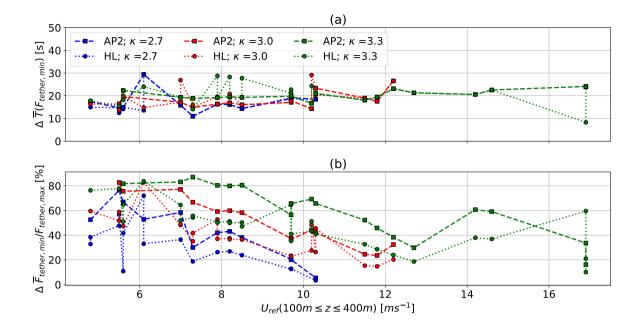


Figure 9. Average time between tether tension troughs (a) and relative decrease in tether tension (b) during the production phase of optimized ground-generation AWES with a wing area of $A_{wing} = 50 \text{ m}^2$. The figures diagram show results for mass scaling exponents $\kappa = 2.7, 3.0, 3.3$ (blue, red, green) and both sets of aerodynamic coefficients AP2 reference (square, dashed line) and high-lift HL (circle, dashed lines).

good proxy for conditions at operating height. Lighter aircraft and higher lift wings results in slightly higher operating altitudes, a longer tether and higher elevation angle.

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Outliers, e.g. for high wind speed profiles (compare Figure 2), are likely local optima of the highly nonlinear trajectory optimization problem described in Section 3.

As wind speed increases beyond rated power ($U_{ref} \approx 10 \text{ ms}^{-1}$, see fFigures 6 and 8), the aircraft moves out of the wind window to de-power. This is seen as indicated by rising average elevation angles $\overline{\epsilon}$ (bottom) above $U_{ref} = 10 \text{ ms}^{-1}$. Results for both offshore (right) and onshore (left) follow the same trends, but operating heights below rated wind speed are lower offshore because of lower wind shear and higher wind speeds.

It is important to keep in mind that even though the operating height exceeds 500 m for wind speeds of more than $U_{\rm ref} \approx 15 \text{ ms}^{-1}$ such wind speeds occur only about 10 % of the time (see Figure 2). These high wind speeds can significantly contribute to AEP. In case of the $A = 20 \text{ m}^2$ wing, this contribution is about 29 % onshore and 33% offshore. Between 5 and 15 ms⁻¹, which is the most likely wind speed range, operating heights both onshore and offshore are between 200 to 300 m.

385 For smaller system sizes these heights are even lower. While this is slightly above the hub-height of current conventional wind turbines, it rebuts the argument of harvesting wind energy beyond this altitude (Archer and Caldeira, 2009). These findings are consistent with current offshore WT trends, whose rotor diameter increased significantly while hub height only increased marginally over the last years. However, It is likely that offshore hub heights will increase as technology improves, making the

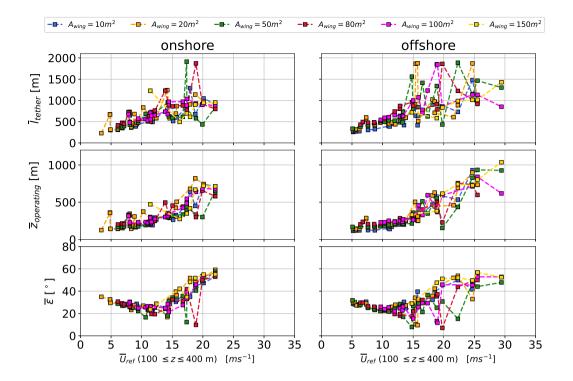


Figure 10. Average tether length \bar{t}_{tether} (top), average operating altitude $\overline{z}_{\text{operating}}$ (center middle) and average elevation angle \overline{z} (bottom) over reference wind speed $\overline{U}_{\text{ref}}(100 \le z \le 400 \text{ m})$. Results for wing areas between $A_{\text{wing}} = 10 - 150 \text{ m}^2$ scaled with an exponent of $\kappa = 3$, AP2 reference aerodynamic coefficients for both onshore (left) and offshore (right) location.

argument for the deployment of AWES particularity challenging as both operate at comparable heights and WT are the more
 proven and established technology. However, this might be different for multiple kite systems which could benefit from longer
 tethers, due to reduced tether motion (De Schutter et al., 2019). Furthermore, the radical mass savings potential of AWES could
 prove a decisive factor to pursue the development.

4.3 Power curve, annual energy distribution and power harvesting factor

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This sub-section In the following, we compares average cycle power \overline{P} , annual energy production distribution E and power harvesting factor ζ (Equation (8)) of optimized trajectories to the quasi steady-state model (QSS QSM) described in Sub-section 3.2.

The top Sub-figures (a) of Figure 12 and 13 Figures 12 (a) and 13 (a) compare the effect of aerodynamic efficiency and location on eyele-average power average cycle power in the form of a power curve for AWESs with a wing area of $A_{\text{wing}} = 50 \text{ m}^2$ and a mass scaling exponent of $\kappa = 2.7$. The data are derived from 3 three representative profiles from each of the 10 wind velocity clusters. The average wind speed between 100 and 400 m has been chosen as reference wind speed, because these

are typical operating heights for these AWES. As of now, no standard reference wind speed U_{ref} , equivalent to wind speed at

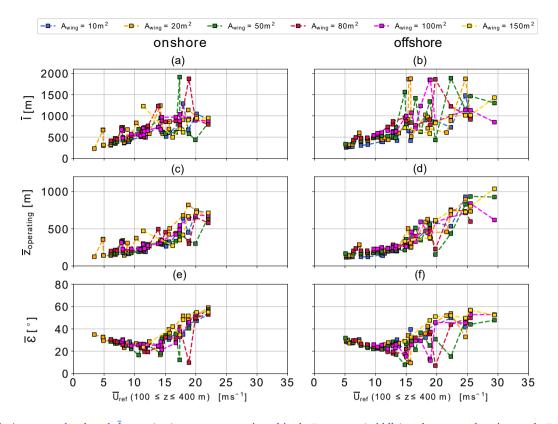


Figure 11. Average tether length $\overline{l}_{\text{tether}}$ (top), average operating altitude $\overline{z}_{\text{operating}}$ (middle) and average elevation angle $\overline{\varepsilon}$ (bottom) over reference wind speed $\overline{U}_{\text{ref}}(100 \le z \le 400 \text{ m})$. Results for wing areas between $A = 10 - 150 \text{ m}^2$ scaled with an exponent of $\kappa = 3$, AP2 reference aerodynamic coefficients for both onshore (left) and offshore (right) location.

hub height for conventional WT, has been agreed upon for AWES. Furthermore, Using this altitude range results in comparable power curve trends onshore and offshore. Offshore AWES could benefit from a larger tether diameter as wind speeds are generally higher (see Figure 2) which would result in higher rated power. -and AEP. Higher lift coefficients result in higher rated power and a steeper power increase up to rated power. Power variations are caused by local optima mostly occurring above rated wind speed as the system de-powers to stay within tether force and flight speed constraints (see Subsection 3.7).

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Rated power, here defined as the maximum, almost constant average cycle power, P_{rated} , defined as the maximum generated power, which is constrained by instantaneous tether force and reeling speed, is summarized in Table 2. Tether reel-in and reel-out speed constraints are kept constant for all designs, simulating drum speed constraints. Tether diameter is kept constant

410 for both locations, but adjusted according to aircraft wing area and aerodynamic efficiency so that all system sizes reach rated power at about $U_{ref} = 10 \text{ms}^{-1}$ (see Subsection 3.5). Therefore, the HL configuration achieves higher rated power. No cut-out wind speed limitations were are implemented. Roland: Aren't these implied by the machine constraints (max. tether force and max. reeling speed)? Because at some point these constraints can not be uphold anymore, which would be a natural cut-out limit. Markus: no, the optimizer generates feasible trajectories up to maximum wind speed, which is why these

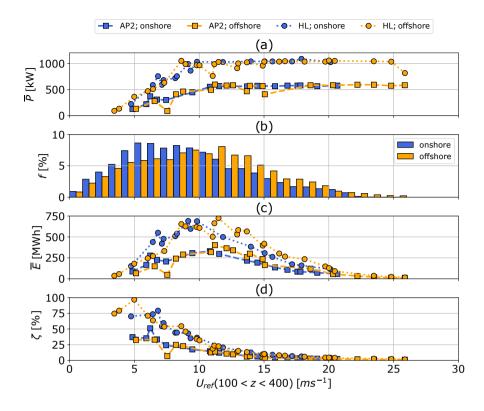


Figure 12. <u>Markus: updated figure</u> Representative AWES power curves (topa) for both sets of HL (circle) and AP2 (square) reference aerodynamic coefficients for both onshore (blue) and offshore (orange) location. The masses of the $A_{wing} = 50 \text{ m}^2$ wing area aircraft are scaled according to Equations (6) and (7) with a mass exponent of $\kappa = 2.7$. Cycle-average power Average cycle power \overline{P} is derived from p5, p50, p95 wind velocity profiles within each of the k = 10 WRF-simulated clusters. A reference height of $100 \le z_{ref} \le 400 \text{ m}$ is used as a proxy for wind speed at operating altitude to calculate the Diagram (b) presents the average annual wind speed probability distribution over reference height range of $100 \le z_{ref} \le 400 \text{ m}$ (center). The integral over the annual Energy production distribution (bottom) which is the product of power and wind speed probability distribution, yields the AEP (legend). The annual energy production distributions over the wind speed are depicted in (c). Diagram (d) shows the corresponding harvesting factor ζ .

- 415 <u>trajectories look so strange and move out of the wind window</u> Therefore, wind power is only limited by each location's maximum wind speed, which is significantly higher offshore (compare Figure 2). Table 2 also shows the estimated equivalent WT rotor diameter D_{WT}^{equiv} , for an assumed power coefficient of $c_p^{WT} = 0.4$ and a rated wind speed of 10ms^{-1} . The system size and therefore material cost benefits of AWES become obvious when comparing AWES wing span b_{wing} to WT rotor diameter D_{WT}^{equiv} . AWES wing span is about 30 (HL) to 40 % (AP2) of the equivalent rotor diameter.
- 420 AEP and cf almost doubles for HL in comparison to the AP2 reference, highlighting the importance of exploring high-lift configurations. The QSS QSM modeled power curves (see Figure 13), which use the same wind velocity profiles and tether diameter as the optimization model, achieve rated power at around $U_{\text{rated}}(100 < z_{\text{ref}} < 400) \approx 8ms^{-1}$. This is caused by the fact that the engineering model neglects mass and predicts optimal power production, whereas the dynamic optimization

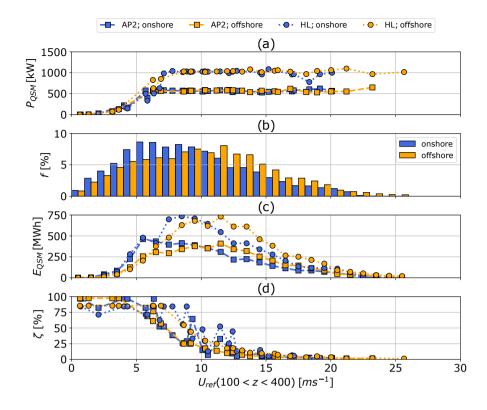


Figure 13. Markus: updated figureQSS model QSM based AWES power curves (topa) for an AWES wing area of $A_{wing} = 50 \text{ m}^2$, both sets of HL (circle) and AP2 (square) reference aerodynamic coefficients and both onshore (blue) and offshore (orange) location. Optimal power \overline{P}_{Loyd} is derived from p5, p25, p50, p75, p95 wind speed profiles within each of the k = 10 WRF-simulated clusters. Diagram (b) presents the average annual wind speed probability distribution over reference height range of $100 \le z_{ref} \le 400 \text{ m}$. The annual energy production distributions over the wind speed are depicted in (c). Diagram (d) shows the corresponding harvesting factor ζ . A reference height of $100 \le z_{ref} \le 400 \text{ m}$ is used as a proxy for wind speed at operating altitude to calculate the annual wind speed probability distribution (centerb). The integral over the annual energy production distribution (bottom) which is the product of power and wind speed probability distribution, yields the AEP (legend).

$A_{\underline{\text{wing}}}[\text{m}^2]$	10		20		50		80		100		150	
$b_{ m wing}[{ m m}]$	10		14.1		22.4		28.3		31.6		38.7	
aerodynamic coeff.	AP2	HL	AP2	HL	AP2	HL	AP2	HL	AP2	HL	AP2	HL
$\overline{P}_{\text{rated}} [\text{kW}]$	145	200	265	420	575	1030	1045	1800	1600	2225	2000	3400
$D_{\rm WT}^{\rm equiv}$ [m]	27	32	37	47	55	73	74	97	91	108	102	132

Table 2. Rated power of AWES with a mass scaling exponent of $\kappa = 2.7$ and equivalent wind turbine rotor diameter

model resolves the flight trajectory and the varying forces and power within each production cycle. Deviation between QSS 425 QSM onshore and offshore power are due to variation in wind conditions. The annual wind speed probability distributions f in Figures 12 and 13 (b) represents the average annual wind speed between $100 \le z \le 400$ m which stands in as a proxy for wind at operating altitude (see Section 2). As expected, higher wind speeds are more likely to occur offshore (FINO3) than onshore (Pritzwalk). However, Very high wind speeds above $U_{ref} > 18 - 20 \text{ms}^{-1}$, beyond the cut-off speed of realistic wind energy converters, have a very low chance occurrence at both locations. The resulting annual average energy production distributions \overline{E} (Figures 12 (c) and 13 (c)) reveal a clear difference between the offshore and

430

Estimated energy production distributions $E_{\underline{QSSQSM}}$, <u>AEP and ef</u> (Figure 13 (c)) of the <u>QSS QSM</u> reference model are based on the same wind speed <u>probability</u> distribution as the optimization model. Here the <u>QSS QSM</u> model data has been

interpolated to be compatible with the annual wind speed probability distribution f (Figure 13 (b)). The QSS QSM model

onshore energy potentials. Better wind conditions offshore result in higher AEP and cf (Sub-section 4.4).

- 435 predicts a higher energy production distribution (Figure 13 (c)) up to rated wind speed than the optimization model, because of the lack of a defined cut-in wind speed and a sooner rated power. Beyond rated power, E_{QSSQSM} is similar to optimized results, as predicted power is very similar, except some small variation. This leads to a higher AEP and cf predictions (Figure 14).
- Sub-figure (d) in figures 12 and 13 Figures 12 (d) and 13 (d) presents the power harvesting factor ζ Diehl (2013), which sets 440 cycle-average AWES power \overline{P} in relation to the total wind power of a cross sectional area $\underline{P_{\text{area}}}$ of the same size as a given wing $A_{\underline{\text{wing}}}$. The power harvesting factor decreases steadily for both the optimization and QSS QSM model. The QSS QSM model predicts an almost constant ζ at low wind speeds ($U_{\text{ref}} < 5 \text{ ms}^{-1}$).

4.4 Annual energy production and capacity factor

The previously described power curves \overline{P} (Figures 12 and 13(a)) and annual wind speed probability distributions f (Figures 445 <u>12 and 13(b)</u>) allow the investigation of the annual energy production distribution E (Figures 12 and 13(c)). AEP is derived from the binned average cycle power P_i , its corresponding wind speed probability f_i and the total hours per year. and therefore annual energy production (AEP) as well as capacity factor (cf):

$$AEP = \sum_{i=1}^{k} \left(\overline{P}_{i} f_{i} \right) 8760 \frac{h}{\text{year}}$$
(9)

cf is calculated from AEP and rated system power P_{rated} (Table 2), defined as the maximum average cycle-power.

450 cf =
$$\frac{\text{AEP}}{P_{\text{rated}} 8760 \frac{\text{h}}{\text{vear}}}$$
 (10)

We assume the same wind speed probability distribution for the QSM model as for the optimization model. The left side of Figure 14 compares onshore AEP (a,b) and cf (c,d) Figure 14 compares AEP for all system sizes scaled with a mass scaling exponent of $\kappa = 2.7$ to QSS QSM data. , while the right side compares offshore results Figures 14 (a) and 14 (c) describe onshore conditions, while Figures 14 (b) and 14 (d) describe offshore conditions. AEP increases almost linearly with wing

- 455 area, because power, which is the product of tether force scales linearly with wing area and when keeping the maximum tether reeling speed is kept constant throughout all optimization runs. As expected, HL aerodynamic coefficients (circle) outperform the AP2 reference (square). Offshore (orange) AEP and cf is generally higher than onshore (blue) because of the higher likelihood of higher wind speeds. The QSS QSM model predicts higher AEP, because of the previously described differences in power up to rated wind speed (Sub-section 4.3), but follows the same trends. The optimization model predicts lower average
- AEP at A_{wing} = 150 m², due to the high number of infeasible solutions at lower wind speeds. Overall cf (Figure 14 c,d) remains almost unchanged up to A_{wing} = 100 m and sharply declines for A_{wing} = 150 m. Onshore AEP and cf seems to outperform offshore for wing areas larger than 100m². This is likely caused by outliers, or wind velocity profile specific local minima, in the power curve before rated wind speed (v_{rated} = 10 ms⁻¹), where the system seemingly over-performs. The QSS QSM model predicts very high cf values at both locations, while offshore AEP always outperforms onshore AEP. The relatively
 high cf values are the result of relativity low rated wind speed. This location_specific design trade-off between generator size,

wing area and tether diameter needs to be further investigated.

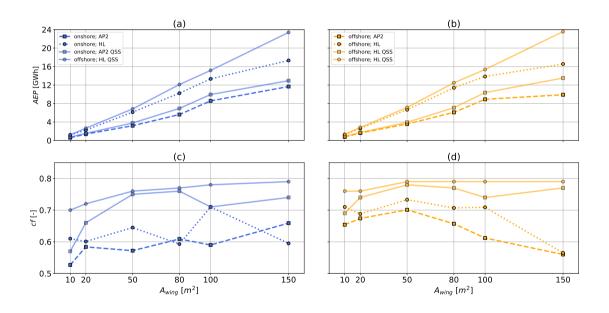


Figure 14. Representative AWES AEP (a,b) and cf (c,d) over as a function of aircraft wing area A_{wing} scaled according to Equations (6) and (7) with a mass exponent of $\kappa = 2.7$. QSS model QSM (solid lines) results are included for reference (Sub-section 3.2). These figures diagrams summarizes data for both sets of HL (circle) and AP2 (square) aerodynamic coefficients as well as both onshore (left, blue) and offshore (right, orange) location. Results are based on the eycle-average power average cycle power \overline{P} derived from p5, p50, p95 wind velocity profiles within each of the k = 10 WRF-simulated clusters and wind speed probability distribution between $100 \le z_{ref} \le 400$ m, used as a proxy for wind speed at operating height.

Figure 15 compares AEP for a mass scaling exponents of $\kappa = 2.7$ to scaling with $\kappa = 3$ and $\kappa = 3.3$, both onshore and offshore. Heavy High mass configurations with no feasible trajectory at any wind speed, which could be interpreted as the wind speed is below the cut-in wind speed, result in missing data. While smaller systems seem to be almost unaffected by

- 470 aircraft weight, mass scaling effects lead to significant reduction in AEP for larger AWES. This is particularly true for wings with the AP2 reference aerodynamic reference coefficients (AP2, square) and onshore wind conditions. Combining results from both Figure 14, which already shows diminishing returns in AEP and cf with increasing wing area for the lightest, idealized aircraft mass scaling, and Figure 15, which predicts that AEP will only decline for heavier mass scaling, conveys that upscaling AWES is only beneficial with significant weight reduction. These results hint at the existence of an upper limit of fixed-wing
- 475 AWES weight relative to AWES size or lift (see Subsection 4.5), which is plausible since mass scales with aircraft volume, assuming pure geometric scaling according to the square-cube law, and lift scales with aircraft area. The mass of soft wing AWES, which are hollow tensile structures filled with air, scales to a great extent with the wing surface, leading to significantly lower mass scaling exponents and more beneficial mass scaling. Therefore and for compensating To account for these scaling effects and considering the power fluctuation caused by the cyclic nature of ground-generation AWES, it is likely better to
- 480 deploy multiple smaller scale devices rather than a single large-scale system. The ideal, site-specific AWES system size needs to be determined by realistic, achievable mass scaling and the local wind resource.

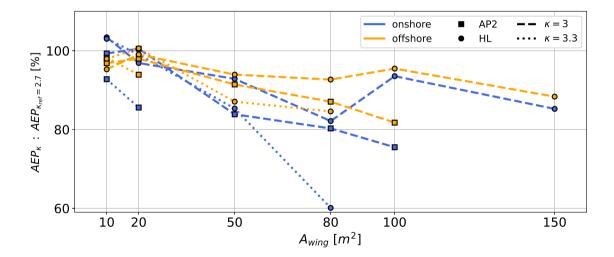


Figure 15. AEP ratio for mass scaling exponent $\kappa = 3$ (dashed lines) and $\kappa = 3.3$ (dotted lines) relative to AEP of $\kappa = 2.7$ over as a function of aircraft wing area A_{wing} . Figure The diagram summarizes data for both onshore (blue) and offshore (orange) locations as well as both sets of aerodynamic coefficients HL (circle) and AP2 (square). Results are based on the eyele-average power average cycle power \overline{P} derived from p5, p50, p95 wind velocity profiles within each of the k = 10 WRF-simulated clusters. Missing data points indicate that no feasible solution for any wind velocity profile was found.

4.5 Impact of weight and drag

The ground-generation crosswind AWES concept benefits from exploits the increased apparent wind speed generated by the flight motion of the thetherd aircraft (Loyd, 1980). Such trajectories, whether circular or figure-of-eight, always include an

485 ascent during every loop <u>maneuver</u> where the aircraft needs to overcome gravity to gain altitude. This leads to a deceleration and therefore reduction of aerodynamic lift. AWESs with excess mass fail to overcome weight and drag and can no longer climb during these phases.

With an increased wing area, the entire aircraft, particularly the load-carrying structures such as the wing box, need to increase in size and weight in order to withstand <u>the</u> increased aerodynamic loads produced by a larger wing. Aircraft mass

- 490 scales with volume ($\kappa = 3$), assuming pure geometric scaling without design improvements, while lift force and therefore power only scales with the area ($F_{\text{lift}} \sim b^2$). However, When tether drag is considered AWES power scales faster than b^2 , because tether drag losses are proportional to the tether diameter, which scales relative to the square root of the wing area. Similarly, conventional WT power and AEP scales with the rotor diameter square, while theoretic WT mass scales with the cube of the rotor diameter. However, Comparing both wind energy converters under these assumptions, AWES perform worse
- 495 with size as their flight path degrades. This can be attributed to the fact that AWES need to produce enough lift to carry their own weight to maintain operational, while WT are supported by a tower.

These facts limit AWES size, as the prevailing wind resource does not improve enough to produce sufficient aerodynamic lift to overcome the increased system drag and weight. An increase of operating altitude only comes with a marginal wind speed increase especially offshore (compare Figure 2). Furthermore, higher operating altitudes also lead to increased cosine losses, unless offset by a longer tether which in turn results in more drag and weight. Better aerodynamics or lighter, more durable

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aircraft and tether materials can only push this boundary, but not overcome it.

A comparison of tether weight W_{tether} during the production phase (reel-out) to total system weight (W_{total} = W_{aircraft} + W_{tether}) in Figure 17 (top) shows that the tether makes up 10 to 30 % of the entire <u>airborne</u> system weight. Note that the tether cross sectional area is sized with a safety factor of 3. Tether cross sectional area mostly scales with aerodynamic force and therefore
505 wing area while the aircraft weight scales with a mass scaling exponent κ > 2.7, 3.0 which results in decreasing trend lines. This value is higher for high-lift airfoils (circle) as the tether diameter is larger to withstand higher aerodynamic forces. For lighter aircraft, scaled with κ = 2.7 (dash-dotted), the portion of tether weight is higher because the tether diameter remains constant while the aircraft mass is lighter.

The bottom sub-figure Figure 17 (bottom) reveals that tether drag makes up about 18 to 40 % of the entire <u>airborne</u> system 510 drag during the production phase. Tether diameter $d_{\underline{tether}}$ and therefore face <u>drag</u> area $(A_{\underline{tether}}^{\underline{facedrag}} = d_{\underline{tether}} \ l_{\underline{tether}})$ scales beneficially with wing area, leading to the downward trend lines. As descried in Sub section 3.5, the total tether drag is divided up evenly between the top and bottom node at every tether segment, resulting in half the tether drag being attributed to the aircraft and the other half to the ground station. Implementing more detailed tether models could explore the impact of tether dynamics and tether drag in more detail.

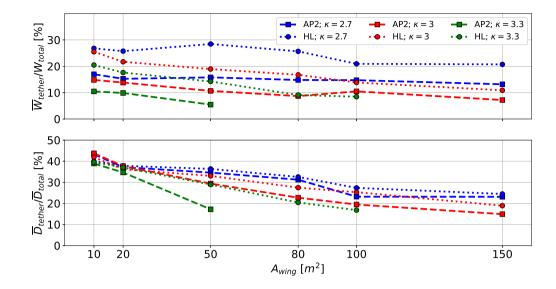


Figure 16. –Percentage of cycle-average tether weight $\overline{W}_{\text{tether}}$ to total weight $\overline{W}_{\text{total}}$ (top) and tether drag $\overline{D}_{\text{tether}}$ to total drag $\overline{D}_{\text{total}}$ (bottom) during production phase (recl-out) for all aircraft sizes $A_{\text{wing}} = 10 - 150 \text{ m}^2$, sets of aerodynamic coefficients AP2, HL and mass scaling exponents $\kappa = 2.7, 3, 3.3$ for wind data at the offshore location.

- 515 It is critical for crosswind AWES to ascend during each loop of the production or reel-out phase. The aircraft needs to produce enough aerodynamic lift, which decreases as the aircraft slows down during ascent (compare Figure 8), to overcome gravity. During the ascent, The aircraft increases angle of attack (Figure 8) to compensate for the decreased apparent wind speed. However, this is not enough to maintain aerodynamic force and tether tension during times of lower wind speeds, especially for larger and heavier systems. For larger and heavier systems, this is not enough to maintain aerodynamic force 520 and tether tension during times of lower wind speeds. The top sub-figure in Figure 18 (a) contrasts shows the aeronautic load factor during the production phase. It is defined as the ratio of average pattern trajectory lift force \overline{L}_{wing} to total AWES weight \overline{W}_{total} , including tether which is sized with a safety factor of 3 and aircraft. , during the reel-out phase for all aircraft sizes $A_{\text{wing}} = 10 - 150 \text{m}^2$, mass scaling exponents $\kappa = 2.7, 3.0, 3.3$ and aerodynamic coefficients HL, AP2. The average load factor decreases from about 10 - 20 to $\frac{10}{5} - 10$, depending on aerodynamic performance and mass scaling, which is approximately the maneuvering load factor of an acrobatic airplane $n_{\text{acrobatic}} = 6.0$ (Federal Aviation Agency, 2017). For utility airplanes 525 this value is about $n_{\text{acrobatic}} = 4.4$. The AWES aeronautic load factor is relatively high in comparison to unterthered aircraft, because the high lift coefficient, which are designed to maximize traction power, in combination with the high wind speeds during the crosswind motion, lead to very high lift forces. The beneficial effect of better aerodynamics and mass scaling are clearly visible in a lower weight to lift higher lift-to-weight ratio. High system mass with insufficient lift on the other hand
- 530 leads to infeasible solutions and missing data.

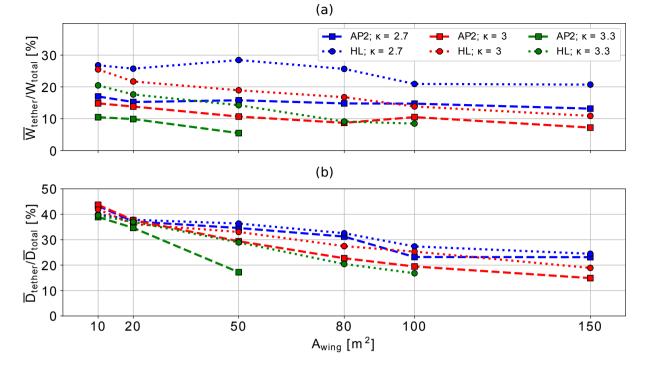


Figure 17. Percentage of cycle-average tether weight $\overline{W}_{\text{tether}}$ to total weight of airborne components $\overline{W}_{\text{total}}$ (a) and tether drag $\overline{D}_{\text{tether}}$ to total drag $\overline{D}_{\text{total}}$ (b) during production phase (reel-out) for all aircraft sizes $A = 10 - 150 \text{ m}^2$, sets of aerodynamic coefficients AP2, HL and mass scaling exponents $\kappa = 2.7, 3, 3.3$ for wind data at the offshore location.

The bottom sub-figure of Figure 18 (bottom) shows a slight reduction of total average drag \overline{D}_{total} to average lift \overline{L}_{wing} ratio with increasing wing area. Overall however, this ratio remains almost constant between 6 to 8 %. The increase for $A_{wing} = 100,150 \ m^2$, $\kappa = 3$ and AP2 aerodynamics is likely caused by local optimization minima and few feasible wind speed profiles.

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For a large-scale aircraft with an area of $A_{\underline{\text{wing}}} = 150 \text{ m}^2$, scaled with the lightest mass scaling exponent of $\kappa = 2.7$, and AP2 reference aerodynamic coefficients, no feasible solution could be found for low wind speeds $U_{\text{ref}} < 5\text{ms}^{-1}$. This can be seen in Figure 19 which shows the total weight $\overline{W}_{\text{total}}$, including tether and aircraft, divided by average lift $\overline{L}_{\text{wing}}$ for all aircraft sizes with AP2 reference aerodynamic scaled with $\kappa = 2.7$. weight-to-lift ratio increases up to $U_{\text{ref}} \approx 5\text{ms}^{-1}$, above which it remains almost constant. This can likely be attributed to the applied apparent flight speed constraint of $U_{\text{app}}^{\text{max}} = 80\text{ms}^{-1}$ which seems to already be achieved at this reference wind speed.

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From this, together with time series data shown in Figure 8, it is possible to to estimate the minimum cut-in wind speed or minimum viable aerodynamic load factor (lift to weight ratio). For the investigated design and constraints, the minimum viable aerodynamic load factor seems to be about 5 which is equivalent to a maximum viable weight_to_lift ratio of 20 %. No feasible solutions were found for lower wind speeds.

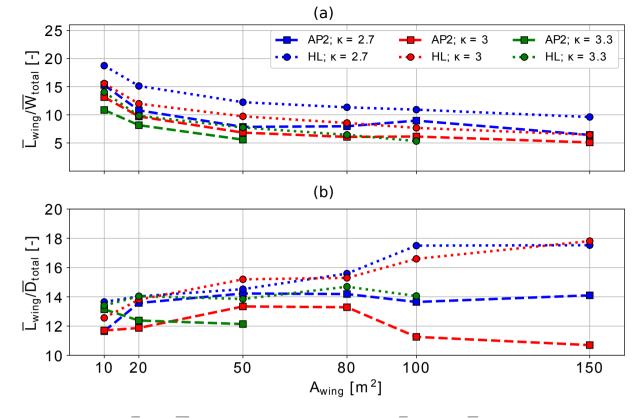


Figure 18. Load factor or lift \overline{L}_{wing} to \overline{W}_{total} ratio (topa) and cycle-average total lift \overline{L}_{wing} to drag \overline{D}_{total} , including tether drag, (bottomb) during production phase (reel-out) for all aircraft sizes $A_{\text{wing}} = 10 - 150 \text{ m}^2$, sets of aerodynamic coefficients AP2, HL and mass scaling exponents $\kappa = 2.7, 3, 3.3$ for wind data at the offshore location. Large-scale results for $A_{\text{wing}} = 100, 150 \text{ m}^2$ might be misleading because only high wind speeds result in feasible solutions (compare figures 19).

The bottom sub-figure of Figure 19 bottom shows the total AWES drag \overline{D}_{total} , including tether drag, to lift ratio \overline{L}_{wing} over 545 reference wind speed for all aircraft sizes scaled with $\kappa = 2.7$ and AP2 reference aerodynamic coefficients. Data for all aircraft sizes show a similar trend with the drag proportion doubling from about 5 to 10 % with wind speed. This increase is primarily caused by the longer tether lengths at higher wind speeds (compare Figure 11), which results in more drag and weight. Heavier system weight also leads to an angle of attack α increase as the aircraft needs to produce more lift while staying within the flight speed constraint. 550

4.6 **Power losses**

An increased aircraft wing area not only leads to increased power potential, but is also accompanied by increased tether losses due to weight and drag. Tether mass scales with aircraft wing size because the higher aerodynamic forces require a larger

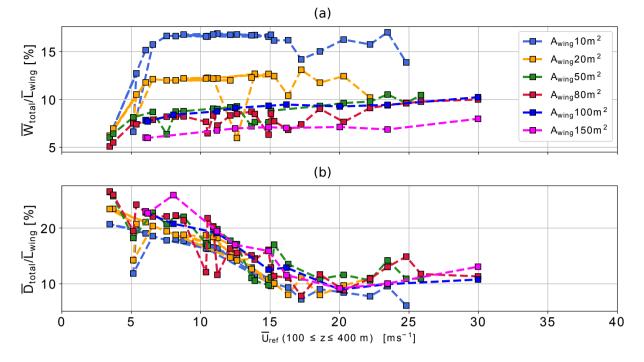


Figure 19. Ratio of cycle-average total weight $\overline{W}_{\text{total}}$ to lift $\overline{L}_{\text{wing}}$ (topa) and cycle-average total drag $\overline{D}_{\text{total}}$, including tether drag, to lift $\overline{L}_{\text{wing}}$ (bottomb) during production phase (reel-out) for all aircraft sizes $A_{\text{wing}} = 10 - 150 \text{ m}^2$ for AP2 reference aerodynamic coefficients and a mass scaling exponent of $\kappa = 2.7$ over reference wind speed offshore.

tether diameter, assuming constant tether strength. Tether length increases with AWES size and wind speed (see Subsection 4.2) which further increases tether drag and weight.

Figure 20 compares the average tether power losses due to the tether $\overline{P}_{tether}^{drag}$, calculated from the tether drag assigned to the aircraft node and its flight speed, relative to average cycle power \overline{P} . for all aircraft wing sizes $A_{wing} = 10 - 150 \text{ m}^2$, mass scaling exponents $\kappa = 2.7, 3.0, 3.3$ and both sets of aerodynamic coefficients (HL, AP2). This power loss can be interpreted as how much of the harvested wind power is dissipated by the tether. Indirect power losses associated with a larger tether, such as the reduction in flight speed due to drag and weight, are not included in this analysis. The relative tether drag loss decreases with wing area, because tether diameter scales beneficially with the square root of the tether force which scales linearly with wing area. This scaling trend is encouraging, but is counteracted and dominated by mass increases with size highlighted in earlier sections. As expected, the high-lift airfoil HL (dotted lines) experiences less relative drag loss than the AP2 reference

airfoil (dashed lines) due to higher average cycle power.

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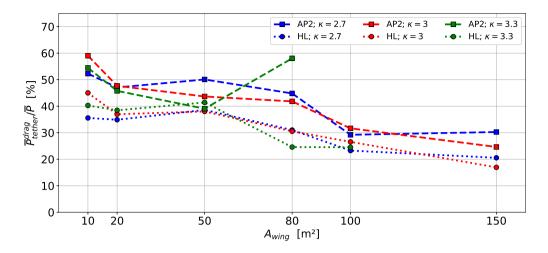


Figure 20. Ratio of eycle-average power average cycle power losses due to tether drag $\overline{P}_{\text{tether}}^{\text{drag}}$ to produced power \overline{P} over aircraft size A_{wing} for both sets of aerodynamic coefficients AP2, HL, all mass scaling exponent of $\kappa = 2.7, 3.0, 3.3$ and wind data at the offshore location.

565 5 Summary and conclusion

This study presentsed rigid wing AWES scaling trends based on the Ampyx AP2 reference and explores the AWES design space subject to representative onshore (Pritzwalk in nothern Germany) and offshore (FINO3 research platform in the North Sea) wind conditions. Generator limitations on speed, torque and power were indirectly implemented by setting a fixed tether reeling speed range and diameter of every design (size and aerodynamic coefficients). This resulted in a constant maximum

- 570 power and a power curve as a function of wind speed. Tether reeling speed range and tether diameter of every design (size and aerodynamic coefficients) were fixed. This can be interpreted as generator limitations on speed, torque and power which result in a constant maximum AWES power and a power curve as a function of wind speed. We evaluated the impact of wing area and mass scaling as well as nonlinear aerodynamic properties on optimal trajectories, reaction forces and moments, power generation and AEP, based on the awebox power and trajectory optimization model. We analyzed the performance for two sets
- 575 of nonlinear aerodynamic coefficients, the AP2 reference and a high-lift configuration where AP2 coefficients were modified as if high-lift devices were attached. Wing areas between $A_{wing} = 10 - 150$ m, with mass properties scaled according to a geometric scaling law with three different mass scaling exponents $\kappa = 2.7, 3.0, 3.3$, were implemented into the awebox power and trajectory optimization toolbox. A representative set of k-means clustered onshore and offshore wind velocity profiles, derived from the mesoscale WRF model, were used to define wind inflow conditions.
- 580 Our analyses estimated a maximum average weight-toaverage-lift ratio (equivalent to an aerodynamic load factor of about 5) to allow viable AWES operation of about 20 %. Our results imply the existence of an upper limit of AWES weight relative to AWES to size or lift, which is plausible since mass scales with aircraft volume and lift scales with aircraft area. In comparison, conventional WT power scales with the square and mass with the cube of the rotor diameter. Under the same

assumptions rigid wing AWES performance scales worse because the aircraft needs to carry the entire increasing system

585 weight (including tether mass), instead of being supported by a tower. Therefore, the optimal AWES size is always defined by the maximum system weight, including tether and aircraft, which the aircraft can support, subject to local wind conditions.

In this work we described and analyzed ground-generation AWES scaling effects subject to realistic wind velocity profiles for a set of representative AWES. We compared the impact of several design parameters based on the Ampyx AP2 reference model. We analyzed AWES performance for two sets of nonlinear aerodynamic coefficients, the AP2 reference and a high-lift

- 590 configuration where AP2 coefficients were adjusted as if high-lift devices were attached. To assess AWES scaling potential, several wing areas between $A_{\underline{\text{wing}}} = 10 - 150$ m, with mass properties scaled according to a geometric scaling law with three different mass scaling exponents $\kappa = 2.7, 3.0, 3.3$, were implemented into the awebox power and trajectory optimization toolbox. A representative set of k-means clustered onshore and offshore wind velocity profiles, derived from the mesoscale WRF model, were used to define wind inflow conditions.
- 595 We discussed the impact of mass and system size on typical trajectories and time series data which confirms that instantaneous power can drop to zero during the reel-out phase. This is caused by insufficient lift as the aircraft tries to overcome gravity ascent and maintain tether tension. The minimum wind speed to sustain positive power production during the reel-out phase as well as tether length and average operating altitude increase with system size and weight. However, operating heights beyond 500 m are rare and mostly occur as the system de-powers above rated wind speed to stay within tether force and flight
- 600 speed constraints. Therefore, it could be reasonable to keep the maximum tether length and operating altitude below those values to reduce costs and permitting burdens. As these constraints become active, the resulting trajectory deforms and diverge from the expected paths seen for lower wind speeds. This is especially true for high-lift configurations as they reach these limits faster. er losses, Since detailed design and stress analysis is beyond the scope of this paper, we limit ourselves to the estimation of maximum mechanical wing loads, assuming an elliptical lift distribution. Shear force scales linearly with wing area, while
- 605 bending moment increases nonlinearly with wing area and wing span.

Analyzing the AWES power curve We determined that rated power scales linearly with wing area, assuming that the tether reeling speed constraints are kept constant and the tether diameter is adjusted appropriately. We chose to size the tether diameter so that rated power is achieved at about $U_{ref} = 10 m s^{-1}$, independent of size, mass and location. A larger tether diameter would increase rated power and shift rated speed towards higher wind speeds, which might be beneficial for faster offshore wind conditions, but would impact tether drag and weight. Improving Higher aerodynamic efficiency increases power production. For the sets of aerodynamic coefficients used in this study, average power increased by approximately 30% to 80 %, for the sets of aerodynamic coefficients used in this study, depending on wing area. We proposed the nonlinear AWES power coefficient $\frac{AWES}{p} \frac{e_{wing}^{-1}}{p}$ from the aircraft path length and wing area to collapse power curves.

We estimated AEP and cf based on the power curve analysis and wind speed probability distribution at reference height 615 between $100 \le z_{ref} \le 400$ m. Offshore AEP is were generally higher than onshore, while the power curves are almost identical even though clustered profiles differ, due to higher wind speeds. Increased aircraft mass leads to <u>a</u> significant reduction in AEP, as lower wind speeds become infeasible to fly in until finally no feasible solutions, even at higher wind speeds, <u>can be</u> was found. This is particularly true for the onshore location and AP2 reference aerodynamics, as these conditions can not produce sufficient lift force to overcome system weight. Wind farm setups might therefore benefit from the deployment of

620 multiple smaller AWES rather than few large-scale AWES. This could also reduce the overall power loss when synchronizing phase-shifting the flight trajectories of AWESs within a farm. Determining the ideal, site-specific AWES size needs to be determined subject to realistic mass scaling, the available area and the local wind resource.

Furthermore, we described the tether contribution to total weight and drag relative to aircraft wing size as well as tetherassociated power losses. Our results show that even though relative tether power losses decrease with wing size, they still use up a significant portion (20 - 60 %) of the average mechanical AWES power.

Lastly, we try to determine investigated the maximum AWES weight_to_lift ratio. Our data showsed that total AWES weight, including tether and aircraft, should not exceed 20 % of the produced aerodynamic lift to operate. The limitation of crosswind AWES operations seems to be the upward climb within each loop. During this ascent the aircraft decelerates by approximately 20%- 25%, which reduces aerodynamic lift by about 35% - 45%, which could be offset by the deployment of additional high-

630 lift devices. As a result, the system can not produce enough lift to overcome gravity and maintain tether tension, leading to a reduction in tether reeling speed and produced power up until a complete drop to zero for lower wind speeds. In comparison, conventional WT power scales with the square and mass with the cube of the rotor diameter. Under the same assumptions rigid wing AWES performance scales worse because the aircraft needs to carry the entire increasing system weight (including tether mass), instead of being supported by a tower. Therefore, the optimal AWES size is always defined by the maximum system
635 weight, including tether and aircraft, which the aircraft can support, subject to local wind conditions.

6 Future work

Defining It is crucial to investigate the AWES design space subject to realistic wind conditions and operating constraints is erucial for scaling to further the development of this technology for large-scale deployment of grid-integration. We therefore propose to build upon this study and further investigate the design space using design optimization. A possible approach is to utilize the already existing AWES power and trajectory optimization toolbox awebox and implement it into a design optimization framework that varies parameters such as aspect ratio, wing area and wing box dimensions. Adding a cost model would allow to optimize for levelized cost of electricity or AEP. Analyzing the dynamic aircraft wing loads caused by the cyclic nature of crosswind AWES and turbulence could improve AWES durability and further explore AWES design by considering fatigue loads to explore wing concepts to minimize κ. Ultimately, AWES must compete with conventional wind. Scaling and moving offshore are logical goals for both technologies. The relative merits of large-scale AWES must be further explored to set design and development targets, particularly since this study highlighted that offshore AWES are not particularly beneficial

relative to conventional wind, given the generally lower sheer offshore. This particularly applies to offshore, where they are in direct competition with WTs, as they both operate at lower altitudes, given the generally lower wind speed. This further highlights that the advantage of ground-generation AWES, in particularly offshore, does not lie in higher altitudes, but in reduced material and associated benefits such as easier transportation. Data availability. TEXT

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Author contributions. Markus Sommerfeld evaluated the data and wrote the manuscript in consultation and under the supervision of Curran Crawford. Martin Dörenkämper set up the numerical offshore simulation, contributed to the meteorological evaluation of the data. Jochem De Schutter co-developed the optimization model and helped writing and reviewed this manuscript.

Competing interests. We declare that neither the author nor any co-author can identify any conflict of interest regarding the subject matter or materials discussed in this manuscript.

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References

2021.

Airborne Wind Europe: Airborne Wind Energy Glossary, https://airbornewindeurope.org/resources/glossary-2/, last accessed: 2022-03-29,

670

685

Ampyx: Ampyx Power BV, https://www.ampyxpower.com/, 2020.

- Andersson, J., Åkesson, J., and Diehl, M.: CasADi: A Symbolic Package for Automatic Differentiation and Optimal Control, in: Recent Advances in Algorithmic Differentiation, edited by Forth, S., Hovland, P., Phipps, E., Utke, J., and Walther, A., pp. 297–307, Springer Berlin Heidelberg, Berlin, Heidelberg, 2012.
- 675 Andersson, J. A. E., Gillis, J., Horn, G., Rawlings, J. B., and Diehl, M.: CasADi A software framework for nonlinear optimization and optimal control, Mathematical Programming Computation, 11, 1–36, https://doi.org/10.1007/s12532-018-0139-4, 2019.
 - Archer, C. L. and Caldeira, K.: Global Assessment of High-Altitude Wind Power, Energies, 2, 307–319, https://doi.org/10.3390/en20200307, 2009.
 - Argatov, I. and Silvennoinen, R.: Efficiency of Traction Power Conversion Based on Crosswind Motion, in: Airborne Wind Energy, edited
- by Ahrens, U., Diehl, M., and Schmehl, R., pp. 65–79, Springer Berlin Heidelberg, Berlin, Heidelberg, https://doi.org/10.1007/978-3-642-39965-7_4, 2013.
 - Argatov, I., Rautakorpi, P., and Silvennoinen, R.: Estimation of the mechanical energy output of the kite wind generator, Renewable Energy, 34, 1525–1532, https://doi.org/10.1016/j.renene.2008.11.001, http://linkinghub.elsevier.com/retrieve/pii/S096014810800400X, 2009.

Bronnenmeyer, T.: Optimal Control for Multi-Kite Emergency Trajectories, Master's thesis, University of Stuttgart, https://cdn.syscop.de/ publications/Bronnenmeyer2018.pdf, 2018.

- Champion, K. S. W., Cole, A. E., and Kantor, A. J.: Chapter 14 Standard and Reference Atmospheres, in: Handbook of Geophysics and the Space Environment., edited by Jursa, A. S., Air Force Geophysics Lab., Hanscom AFB, MA (USA), https://www.osti.gov/biblio/5572212, 1985.
- De Schutter, J., Leuthold, R., Bronnenmeyer, T., Paelinck, R., and Diehl, M.: Optimal control of stacked multi-kite systems for utility-scale airborne wind energy, in: 2019 IEEE 58th Conference on Decision and Control (CDC), pp. 4865–4870, 2019.
 - De Schutter, J., Malz, E., Leuthold, R., Bronnenmeyer, T., Paelinck, R., and Diehl, M.: awebox: Modelling and optimal control of singleand multiple-kite systems for airborne wind energy, https://github.com/awebox, last accessed: 2021-10-11, 2020.
 - Diehl, M.: Airborne Wind Energy: Basic Concepts and Physical Foundations, in: Airborne Wind Energy, edited by Ahrens, U., Diehl, M., and Schmehl, R., pp. 3–22, Springer Berlin Heidelberg, https://doi.org/10.1007/978-3-642-39965-7_1, 2013.
- 695 Dörenkämper, M., Olsen, B. T., Witha, B., Hahmann, A. N., Davis, N. N., Barcons, J., Ezber, Y., García-Bustamante, E., González-Rouco, J. F., Navarro, J., Sastre-Marugán, M., Sīle, T., Trei, W., Žagar, M., Badger, J., Gottschall, J., Sanz Rodrigo, J., and Mann, J.: The Making of the New European Wind Atlas – Part 2: Production and Evaluation, Geosci. Model Dev. Discuss., 2020, 1–37, https://doi.org/10.5194/gmd-2020-23, 2020.

Drela, M. and Youngren, H.: AVL, http://web.mit.edu/drela/Public/web/avl/, last accessed: 2022-03-14, 2016.

- 700 Echeverri, P., Fricke, T., Homsy, G., and Tucker, N.: The Energy Kite Selected Results From the Design, Development and Testing of Makani's Airborne Wind Turbines - Part 1, Technical Report, Makani Power, https://storage.googleapis.com/x-prod.appspot.com/files/ Makani_TheEnergyKiteReport_Part1.pdf, 2020.
 - Eijkelhof, D., Rapp, S., Fasel, U., Gaunaa, M., and Schmehl, R.: Reference Design and Simulation Framework of a Multi-Megawatt Airborne Wind Energy System, Journal of Physics: Conference Series, 1618, https://doi.org/10.1088/1742-6596/1618/3/032020, 2020.

- 705 Ellis, G. and Ferraro, G.: The Social Acceptance of Wind Energy: Where we stand and the path ahead., EUR Scientific and Technical Research Reports, European Commission, https://doi.org/10.2789/696070, 2016.
 - Faggiani, P. and Schmehl, R.: Design and Economics of a Pumping Kite Wind Park, in: Airborne Wind Energy: Advances in Technology Development and Research, edited by Schmehl, R., pp. 391–411, Springer Singapore, Singapore, https://doi.org/10.1007/978-981-10-1947-0_16, https://doi.org/10.1007/978-981-10-1947-0_16, 2018.
- 710 Federal Aviation Agency: Part 23 Airworthiness Standards: Normal, Utility, Acrobatic, and Commuter Category Airplanes, https://www.govinfo.gov/content/pkg/CFR-2017-title14-vol1/xml/CFR-2017-title14-vol1-part23.xml#seqnum23.337, 2017.
 - Fingersh, L., Hand, M., and Laxson, A.: Wind Turbine Design Cost and Scaling Model, Tech. rep., NREL, https://doi.org/10.2172/897434, OSTI Identifier: 897434, 2006.

Gros, S., Zanon, M., and Diehl, M.: A relaxation strategy for the optimization of airborne wind energy systems, in: Control Conference (ECC), 2013 European, pp. 1011–1016, IEEE, https://doi.org/10.23919/ECC.2013.6669670, 2013.

- Haas, T., Schutter, J. D., Diehl, M., and Meyers, J.: Wake characteristics of pumping mode airborne wind energy systems, Journal of Physics: Conference Series, 1256, 012 016, https://doi.org/10.1088/1742-6596/1256/1/012016, 2019.
 - Hoerner, S.: Fluid-dynamic Drag: Practical Information on Aerodynamic Drag and Hydrodynamic Resistance, Hoerner Fluid Dynamics, https://hoernerfluiddynamics.com/, 1965.
- 720 Horn, G., Gros, S., and Diehl, M.: Numerical Trajectory Optimization for Airborne Wind Energy Systems Described by High Fidelity Aircraft Models, in: Airborne Wind Energy, edited by Ahrens, U., Diehl, M., and Schmehl, R., pp. 205–218, Springer Berlin Heidelberg, Berlin, Heidelberg, https://doi.org/10.1007/978-3-642-39965-7_11, 2013.
 - Houska, B. and Diehl, M.: Optimal control for power generating kites, in: 2007 European Control Conference (ECC), pp. 3560–3567, IEEE, https://doi.org/10.23919/ECC.2007.7068861, 2007.
- 725 HSL: HSL. A collection of Fortran codes for large scale scientific computation, http://www.hsl.rl.ac.uk/, last accessed: 2022-03-14, 2020. Hurt, H. J.: Aerodynamics for naval aviators, Naval Air Systems Command, https://www.faa.gov/regulations_policies/handbooks_manuals/ aviation/media/00-80t-80.pdf, 1965.

Kermode, A., Barnard, R., and Philpott, D.: Mechanics of Flight, Pearson Prentice Hall, 2006.

Kitepower B.V.: Kitepower, Online, https://thekitepower.com/, last accessed: 2022-03-17, 2016.

- 730 Kruijff, M. and Ruiterkamp, R.: A Roadmap Towards Airborne Wind Energy in the Utility Sector, in: Airborne Wind Energy: Advances in Technology Development and Research, edited by Schmehl, R., pp. 643–662, Springer Singapore, Singapore, https://doi.org/10.1007/978-981-10-1947-0_26, 2018.
 - Lee, T. and Su, Y. Y.: Lift enhancement and flow structure of airfoil with joint trailing-edge flap and Gurney flap, Experiments in Fluids, 50, 1671–1684, https://doi.org/10.1007/s00348-010-1024-8, 2011.
- 735 Leuthold, R., De Schutter, J., Malz, E., Licitra, G., Gros, S., and Diehl, M.: Operational Regions of a Multi-Kite AWE System, in: 2018 European Control Conference (ECC), pp. 52–57, https://doi.org/10.23919/ECC.2018.8550199, 2018.
 - Licitra, G.: Identification and optimization of an airborne wind energy system, Ph.D. thesis, University of Freiburg, https://doi.org/10.6094/UNIFR/16226, 2018.

Loyd, M. L.: Crosswind kite power (for large-scale wind power production), Journal of Energy, 4, 106–111, https://doi.org/10.2514/3.48021,

740 https://doi.org/10.2514/3.48021, 1980.

715

Luchsinger, R. H.: Pumping Cycle Kite Power, in: Airborne Wind Energy, edited by Ahrens, U., Diehl, M., and Schmehl, R., pp. 47–64, Springer, Berlin, Heidelberg, https://doi.org/10.1007/978-3-642-39965-7_3, 2013.

- Lunney, E., Ban, M., Duic, N., and Foley, A.: A state-of-the-art review and feasibility analysis of high altitude wind power in Northern Ireland, Renewable and Sustainable Energy Reviews, 68, 899 911, https://doi.org/https://doi.org/10.1016/j.rser.2016.08.014, 2017.
- 745 Malz, E., Koenemann, J., Sieberling, S., and Gros, S.: A reference model for airborne wind energy systems for optimization and control, Renewable Energy, 140, 1004 – 1011, https://doi.org/10.1016/j.renene.2019.03.111, 2019.
 - Malz, E., Verendel, V., and Gros, S.: Computing the power profiles for an Airborne Wind Energy system based on large-scale wind data, Renewable Energy, 162, 766 – 778, https://doi.org/https://doi.org/10.1016/j.renene.2020.06.056, 2020.
- Malz, E. C., Zanon, M., and Gros, S.: A Quantification of the Performance Loss of Power Averaging in Airborne Wind Energy Farms, in: 2018 European Control Conference (ECC), pp. 58–63, https://doi.org/10.23919/ECC.2018.8550357, 2018.
- North, A., Siegwart, R., and Engel, W.: Autonomous Solar UAV for Sustainable Flights, in: Advances in Unmanned Aerial Vehicles: State of the Art and the Road to Autonomy, edited by Valavanis, K. P., pp. 377–405, Springer Netherlands, Dordrecht, https://doi.org/10.1007/978-1-4020-6114-1_12, 2007.
- offshorewind.biz: Ampyx Power Taking First Steps Towards MW-Scale Kite, Online, https://www.offshorewind.biz/2018/12/04/ 755 ampyx-power-taking-first-steps-towards-mw-scale-kite/, last accessed: 2022-03-14, 2018.
- Pedregosa, F., Varoquaux, G., Gramfort, A., Michel, V., Thirion, B., Grisel, O., Blondel, M., Prettenhofer, P., Weiss, R., Dubourg, V., Vanderplas, J., Passos, A., Cournapeau, D., Brucher, M., Perrot, M., and Duchesnay, E.: Scikit-learn: Machine Learning in Python, Journal of Machine Learning Research, 12, 2825–2830, https://doi.org/10.48550/arXiv.1201.0490, 2011.
- Salvação, N. and Guedes Soares, C.: Wind resource assessment offshore the Atlantic Iberian coast with the WRF model, Energy, 145, 276 –
 287, https://doi.org/https://doi.org/10.1016/j.energy.2017.12.101, 2018.
 - Schelbergen, M., Kalverla, P. C., Schmehl, R., and Watson, S. J.: Clustering wind profile shapes to estimate airborne wind energy production, Wind Energy Science, 5, 1097–1120, https://doi.org/10.5194/wes-5-1097-2020, 2020.
 - Schmehl, R., Noom, M., and van der Vlugt, R.: Traction Power Generation with Tethered Wings, in: Airborne Wind Energy, edited by Ahrens, U., Diehl, M., and Schmehl, R., chap. 2, pp. 23–45, Springer, Berlin, Heidelberg, https://doi.org/10.1007/978-3-642-39965-7_2, 2013.
 - Scholz, D.: Aircraft Design an Open Educational Resource (OER) for Hamburg Open Online University (HOOU), http://HOOU.ProfScholz. de, last accessed: 2021-10-11, 2016.
 - SkySails Group GmbH: SkySails, Online, https://skysails-group.com/, last accessed: 2022-03-17, 2001.

765

- Sommerfeld, M., Crawford, C., Monahan, A., and Bastigkeit, I.: LiDAR-based characterization of mid-altitude wind conditions for airborne wind energy systems, Wind Energy, 22, 1101–1120, https://doi.org/10.1002/we.2343, 2019a.
- Sommerfeld, M., Dörenkämper, M., Steinfeld, G., and Crawford, C.: Improving mesoscale wind speed forecasts using lidar-based observation nudging for airborne wind energy systems, Wind Energy Science, 4, 563–580, https://doi.org/10.5194/wes-4-563-2019, 2019b.
 - Sommerfeld, M., Dörenkämper, M., De Schutter, J., and Crawford, C.: Offshore and onshore ground-generation airborne wind energy power curve characterization, Wind Energy Science Discussions, 2020, 1–39, https://doi.org/10.5194/wes-2020-120, 2020.
- 575 Stull, R.: An Introduction to Boundary Layer Meteorology, Atmospheric and Oceanographic Sciences Library, Springer Netherlands, 1988. van der Vlugt, R., Bley, A., Noom, M., and Schmehl, R.: Quasi-steady model of a pumping kite power system, Renewable Energy, 131, 83 – 99, https://doi.org/10.1016/j.renene.2018.07.023, 2019.
 - Vimalakanthan, K., Caboni, M., Schepers, J., Pechenik, E., and Williams, P.: Aerodynamic analysis of Ampyx's airborne wind energy system, Journal of Physics: Conference Series, 1037, 062 008, https://doi.org/10.1088/1742-6596/1037/6/062008, 2018.
- 780 Waechter, A. and Laird, C.: Ipopt (Interior Point OPTimizer), https://github.com/coin-or/Ipopt, last accessed: 2022-03-14, 2016.

Wächter, A. and Biegler, L. T.: On the implementation of an interior-point filter line-search algorithm for large-scale nonlinear programming, Mathematical Programming, 106, 25–57, https://doi.org/10.1007/s10107-004-0559-y, 2006.