



Wind Energy Resource Maps of South Carolina

Prepared for:

South Carolina Energy Office

1201 Main Street, Suite 1010

Columbia, SC 20201

Attention: Richard Horton

Prepared by:

AWS Truewind, LLC

255 Fuller Road, Suite 274

Albany, New York 12203

Telephone: (518) 437-8660

mbrower@awstruewind.com

Principal Author: Rebecca Reed

Reviewer: Michael Brower

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TABLE OF CONTENTS

LIST OF TABLES.....	2
LIST OF FIGURES.....	3
EXECUTIVE SUMMARY.....	4
1. INTRODUCTION.....	5
2. DESCRIPTION OF THE MESOMAP SYSTEM.....	5
3. IMPLEMENTATION OF MESOMAP FOR THIS PROJECT	8
4. VALIDATION	10
5. WIND MAPS	12
6. GUIDELINES FOR USE OF THE MAPS	12

AWS Truewind

LIST OF TABLES

Table 1. Range of Surface Roughness Values for Leading Land Cover Types9
Table 2. Wind Speed Validation.....11



LIST OF FIGURES

Figure 1. Wind Speed Validation.....8



EXECUTIVE SUMMARY

This report describes a wind-mapping project conducted by AWS Truewind for the South Carolina Energy Office. Using the MesoMap system, AWS Truewind produced maps of mean annual wind speed in South Carolina for heights of 30, 50, 70, and 100 m above ground, as well as maps of mean annual wind power at 50 and 100 m. AWS also produced data files of the predicted wind speed frequency distribution and speed and energy by direction. The maps and data files are provided on a CD with the ArcReader software, which will enable users to view, print, copy, and query the maps and wind rose data.

The MesoMap system consists of an integrated set of atmospheric simulation models, databases, and computers and storage systems. At the core of MesoMap is MASS (Mesoscale Atmospheric Simulation System), a numerical weather model, which simulates the physics of the atmosphere. MASS is coupled to a simpler wind flow model, WindMap, which is used to refine the spatial resolution of MASS and account for localized effects of terrain and surface roughness. MASS simulates weather conditions over a region for 366 historical days randomly selected from a 15-year period. When the runs are finished, the results are input into WindMap. In this project, the MASS model was run on a grid spacing of 2.5 km and WindMap on a grid spacing of 200 m.

AWS Truewind subsequently validated the wind maps using data from 15 stations. The data were first extrapolated to a height of 50 m. The predicted wind speeds are on average about 0.03 m/s higher than the observed/extrapolated speeds. The standard deviation of the biases is 0.41 m/s, or about 8.6% of the average speed at all the stations. The error margin in meters/second is comparable to that obtained in other MesoMap projects, but in percentage terms it is somewhat larger than usual because of the relatively low average speeds.

The wind maps indicate that the wind resource of South Carolina is relatively good offshore and at exposed points along the coast but declines substantially inland. Well offshore, the mean wind speed at 70 m height ranges from 8.0 to 8.5 m/s, and the wind power density ranges from 600 to 800 W/m² (NREL class 6); nearer the coast, the mean speed at 70 m ranges from 6.5 to 7.5 m/s and the wind power 300 to 400 W/m² (NREL class 3). The mean wind speed at 70 m height in coastal areas and inland lakes is predicted to be 5.5 to 7.0 m/s, and the predicted mean wind power density at 50 m is 200 to 300 W/m², or NREL class 2. The wind resource in the rest of the state is generally less than 6.0 m/s at 70 m and the wind power density less than 200 W/m² (NREL class 1), except for a few high ridges in the extreme northwestern corner of the state along the North Carolina border, where the mean speed may reach 8 m/s. The wind resource increases significantly with height above ground, however, especially away from the coast. Winds at 100 m are typically 12-15% stronger and contain 40-50% more energy than at 70 m.

1. INTRODUCTION

The South Carolina Energy Office is interested in assessing the potential for wind energy development in South Carolina and assisting developers in finding suitable sites for wind energy projects. Conventional field techniques of wind resource assessment can be time consuming, however, and often depend heavily on local meteorological expertise as well as the availability of reliable and representative wind measurements. Conventional wind flow models, on the other hand, have often proven inaccurate in complex wind regimes, and even in moderate terrain their accuracy can decline substantially with distance from the nearest available reference mast.

Mesoscale-microscale modeling techniques offer a solution to these challenges. By combining a sophisticated numerical weather model capable of simulating large-scale wind patterns with a microscale wind flow model responsive to local terrain and surface conditions, they enable the mapping of wind resources over large regions with much greater accuracy than has been possible in the past. In addition, they do not require surface wind data to make reasonably accurate predictions. While on-site measurements are still required to confirm the predicted wind resource at any particular location, mesoscale-microscale modeling can greatly reduce the time and cost to identify and evaluate potential wind project sites.

AWS Truewind has been the world leader in the development of mesoscale-microscale mapping techniques, having introduced the MesoMap system in the late 1990s. In the past five years, MesoMap has been applied in nearly 30 countries on four continents. In North America alone, MesoMap has been used to map over 30 US states and several provinces of Canada and states of Mexico.

The objective of the current project was to use MesoMap to create high-resolution wind resource maps of South Carolina and to provide wind resource data in a format enabling users to assess potential sites in a GIS. These objectives have been met. In the following sections, we describe the MesoMap system and mapping process in detail; how MesoMap was applied in this project; the validation process and results; the final wind maps and data files; and guidelines for the use of the maps.

2. DESCRIPTION OF THE MESOMAP SYSTEM

The MesoMap system has three main components: models, databases, and computer systems. These components are described below.

2.1. Models

At the core of the MesoMap system is MASS (Mesoscale Atmospheric Simulation System), a numerical weather model that has been developed over the past 20 years by AWS's partner MESO, Inc., both as a research tool and to provide commercial weather forecasting services. MASS simulates the fundamental physics of the atmosphere including conservation of mass, momentum, and energy, as well as the moisture phases, and it contains a turbulent kinetic energy module that accounts for the effects of viscosity

and thermal stability on wind shear. As a dynamical model, MASS simulates the evolution of atmospheric conditions in time steps as short as a few seconds. This creates great computational demands, especially when running at high resolution. Hence MASS is usually coupled to a simpler but much faster program, WindMap, a mass-conserving wind flow model. Depending on the size and complexity of the region and requirements of the client, WindMap is used to improve the spatial resolution of the MASS simulations to account for the local effects of terrain and surface roughness variations.

2.2. Data Sources

The MASS model uses a variety of online, global, geophysical and meteorological databases. The main meteorological inputs are reanalysis data, rawinsonde data, and land surface measurements. The reanalysis database – the most important – is a gridded historical weather data set produced by the US National Centers for Environmental Prediction (NCEP) and National Center for Atmospheric Research (NCAR). The data provide a snapshot of atmospheric conditions around the world at all levels of the atmosphere in intervals of six hours. Along with the rawinsonde and surface data, the reanalysis data establish the initial conditions as well as updated lateral boundary conditions for the MASS runs. The MASS model itself determines the evolution of atmospheric conditions within the region based on the interactions among different elements in the atmosphere and between the atmosphere and the surface. Because the reanalysis data are on a relatively coarse, 200 km grid, MASS is run in several nested grids of successively finer mesh size, each taking as input the output of the previous nest, until the desired grid scale is reached. This is to avoid generating noise at the boundaries that can result from large jumps in grid cell size. The outermost grid typically extends several thousand kilometers.

The main geophysical inputs are elevation, land cover, vegetation greenness (normalized differential vegetation index, or NDVI), soil moisture, and sea-surface temperatures. The global elevation data normally used by MesoMap were produced by the US Geological Survey in a gridded digital elevation model, or DEM, format from a variety of data sources.¹ The US Geological Survey, the University of South Carolina, and the European Commission's Joint Research Centre (JRC) produced the global land cover data in a cooperative project. The land cover classifications are derived from the interpretation of Moderate Imaging Spectroradiometer (MODIS) data collected by satellite. The model translates both land cover and NDVI data into physical parameters such as surface roughness, albedo, and emissivity. The nominal spatial resolution of all of these data sets is 1 km. Thus, the standard output of the MesoMap system is a 1 km gridded wind map. However, much higher resolution maps can be produced where the necessary topographical and land cover data are available. In the United States, the resolution is typically 50 to 200 m.

¹The US Defense Department's high-resolution Digital Terrain Elevation Data set is the principal source for the global 1 km elevation. Gaps in the DTED data set were filled mainly by an analysis of 1:1,000,000 scale elevation contours in the Digital Chart of the World (now called VMAP).

2.3. Computer and Storage Systems

The MesoMap system requires a very powerful set of computers and storage systems to produce wind resource maps at a sufficiently high spatial resolution in a reasonable amount of time. To meet this need AWS Truewind has created a distributed processing network consisting of 94 Pentium II processors and 3 terabytes of hard disk storage. Since each day simulated by a processor is entirely independent of other days, a project can be run on this system up to 94 times faster than would be possible with any single processor. To put it another way, a typical MesoMap project that would take two years to run on a single processor can be completed in just one week.

2.4. The Mapping Process

The MesoMap system creates a wind resource map in several steps. First, the MASS model simulates weather conditions over 366 days selected from a 15-year period. The days are chosen through a stratified random sampling scheme so that each month and season is represented equally in the sample; only the year is randomized. Each simulation generates wind and other weather variables (including temperature, pressure, moisture, turbulent kinetic energy, and heat flux) in three dimensions throughout the model domain, and the information is stored at hourly intervals. When the runs are finished, the results are compiled into summary data files, which are then input into the WindMap program for the final mapping stage. The two main products are usually (1) color-coded maps of mean wind speed and power density at various heights above ground and (2) data files containing wind speed and direction frequency distribution parameters. The maps and data can then be compared with land and ocean surface wind measurements, and if significant discrepancies are observed, adjustments to the wind maps can be made.

2.5. Factors Affecting Accuracy

In our experience, the most important sources of error in the wind resource estimates produced by MesoMap are the following:

- Finite grid scale of the simulations
- Errors in assumed surface properties such as roughness
- Errors in the topographical and land cover data bases

The finite grid scale of the simulations results in a smoothing of terrain features such as mountains and valleys. For example, a mountain ridge that is 2000 m above sea level may appear to the model to be only 1600 m high. Where the flow is forced over the terrain, this smoothing can result in an underestimation of the mean wind speed or power at the ridge top. Where the mountains block the flow, on the other hand, the smoothing can result in an overestimation of the resource as the model understates the blocking effect. The problem of finite grid scale can be solved by increasing the spatial resolution of the simulations, but at a cost in computer processing and storage.

Errors in the topographical and land cover data can obviously affect wind resource estimates. While elevation data are usually reliable, errors in the size and location of terrain features nonetheless occur from time to time. Errors in the land cover data are more common, usually as a result of the misclassification of aerial or satellite imagery. It has been estimated that the global 1 km land cover database used in the MASS simulations is about 70% accurate. Where possible, more accurate and higher resolution land cover databases should be used in the WindMap stage of the mapping process to correct errors introduced in the MASS simulations. In the United States, we use a 30 m gridded Landsat-derived land cover database for this purpose; a similar 250 m database, called Corine, is available for most of Western Europe.

Even if the land cover types are correctly identified, there is uncertainty in the surface properties that should be assigned to each type, and especially the vegetation height and roughness. The forest category, for example, may include many different varieties of trees with varying heights and density, leaf characteristics, and other features that affect surface roughness. Cropland may be virtually devoid of trees and buildings, or it may have many windbreaks. Uncertainties like these can be resolved only by acquiring more information about the area through aerial photography or field observation. However this is not practical when (as in this project) the area being mapped is very large.

3. IMPLEMENTATION OF MESOMAP FOR THIS PROJECT

The standard MesoMap configuration was used in this project. MASS was run on the following nested grids:

- First (outer) grid level: 30 km
- Second (intermediate) grid level: 10 km
- Third (inner) grid level: 2.5 km

The usual geophysical and meteorological inputs were used. The WindMap program adjusted the wind resource estimates to reflect local topography and surface roughness changes on a grid spacing of 200 m. For the topographical data, we used the National Elevation Dataset, a digital terrain model produced on a 30 m grid by the US Geological Survey (USGS). For the land cover, we used the National Land Cover Dataset, which is derived from Landsat imagery. It was also produced by the USGS on a 30 m grid.² Both data sets are of very high quality.

In converting from land cover to surface roughness, the roughness length values shown in Table 1 were assumed. We believe these values to be typical of conditions in states such as South Carolina. However the actual roughness could vary a good deal within each class.

² Information on the National Land Cover Data set can be found at the following web address: <http://landcover.usgs.gov/nationallandcover.html>. Information on the National Elevation Dataset (NED) can be found at <http://edcwww.cr.usgs.gov/products/elevation/ned.html>.

Table 1. Range of Surface Roughness Values for Leading Land Cover Types

Description	Roughness (m)
Cropland	0.03
Grasslands/Herbaceous	0.03
Shrubland	0.05
Deciduous Forest	0.9
Evergreen and Mixed Forest	1.125
Residential Development	0.3
Urban Development	0.75
Herbaceous Wetland	0.2
Woody Wetland	0.66

The roughness is not the only surface property with a direct effect on near-surface wind speeds. Where there is dense vegetation the wind can skim along the vegetation canopy, thereby displacing the flow above the ground and reducing the speed observed at a fixed height above ground. The displacement height is defined as the height at which the wind speed becomes zero in the logarithmic shear formula. The shear formula is as follows:

$$\frac{v_2}{v_1} = \frac{\ln\left(\frac{z_2 - d}{z_0}\right)}{\ln\left(\frac{z_1 - d}{z_0}\right)}$$

Here, d is the displacement height, z_1 and z_2 are two different heights at which the speed v is measured, and z_0 is the surface roughness (generally much less than z_2 and z_1). Note that according to this formula, when $z_2 = d + z_0$, $v_2 = 0$.

The displacement height is usually estimated to be about two-thirds to three-fourths the maximum vegetation height. For this project, we assumed that the displacement height was 10 times the surface roughness length, which was in turn defined to be approximately 7.5% of the vegetation height. For deciduous forests with a roughness length of 0.9 m, this resulted in a displacement height of 9 m.

The effect of displacement height is to reduce the wind speed observed near the ground and to increase the apparent wind shear measured with respect to ground level. It can also reduce the wind speed measured in small clearings, since the ground appears to be in a “hole” at a depth d below the vegetation canopy. The impact of this hole on wind speed diminishes as the clearing becomes large enough for the flow to reach equilibrium with the new effective ground height. As a rule of thumb, the clearing width should be at least 20 times the displacement height for the effect to be negligible at the center of the clearing, but under some conditions the minimum width should be even larger.

4. VALIDATION

The wind resource maps were initially produced without any reference to surface wind measurements. We then validated the wind maps by comparing the predicted speed against data from 15 airport stations.

The validation was carried out in the following steps:

1. Station locations were verified and adjusted, if necessary, by comparing the quoted elevations and station descriptions against the elevation and land cover maps. Where there was an obvious error in position, the station was moved to the nearest point with the correct elevation and surface characteristics.
2. The observed mean speed and power were extrapolated to a common reference height of 50 m using the power law. The shear exponent was estimated from available information about the sites. Assumed shear values ranged from 0.21 to 0.30 with an average of 0.25.³
3. The error margin for each data point was then estimated as a function of two factors: the tower height and the number of years of measurement. The tower height enters the equation because of uncertainty in the wind shear. We assumed an error margin in the shear exponent of 0.04, reflecting significant uncertainty in the ground cover, tree height, buildings, and other factors. The number of years of data affects the uncertainty because winds recorded over a short period may not be representative of long-term conditions. A rule of thumb is that a mean speed based on one year of data will be within 10% of the true long-term mean with 90% confidence. This translates into a standard error of 6% for one year of data. We assumed that the annual mean varies randomly according to a normal distribution, and thus the error margin varies inversely with the square root of the number of years. An additional uncertainty of 3% was added to account for possible variations in the characteristics of anemometers and data loggers.
4. The various uncertainties were then combined in a least-squares sum as follows:

$$(1) e = \sqrt{0.03^2 + \left(\left(\frac{50}{H} \right)^{0.05} - 1 \right)^2 + \left(\frac{0.06}{\sqrt{N}} \right)^2}$$

where H is the height of the anemometer, and N the number of years of measurement. The uncertainty in power (in percentage terms) is assumed to be three times the uncertainty in speed, since the power varies as the cube of the speed.

³ The power shear exponent is assumed to be $3(\alpha-0.02)$, where α is the speed shear exponent. The reason for the reduction in effective shear, compared to assuming that the power goes strictly as the cube of the speed, is that the speed frequency distribution tends to become narrower with height above ground because the shear is often higher under light wind conditions.

- Next, the predicted and measured/extrapolated speed and power were compared, and the map bias (map speed or power minus measured/extrapolated speed or power) was calculated for each point.

Table 2 summarizes the results. The key finding is that the standard deviation of the biases was 0.41 m/s, or 8.6% of the average observed speed.

Table 2. Wind Speed Validation

	Number of Stations	Mean Bias	Standard Deviation of Bias
Speed	15	0.03 m/s (+0.6%)	0.41 m/s (8.6 %)

The scatter plot in Figure 1 compares the predicted and measured-extrapolated wind speeds at 50 m height. The linear trend line is forced through zero. The error bars reflect the uncertainty in the measured/extrapolated long-term wind speed due to short tower heights and short periods of record. They do not include possible errors due to buildings or other obstructions near the masts.

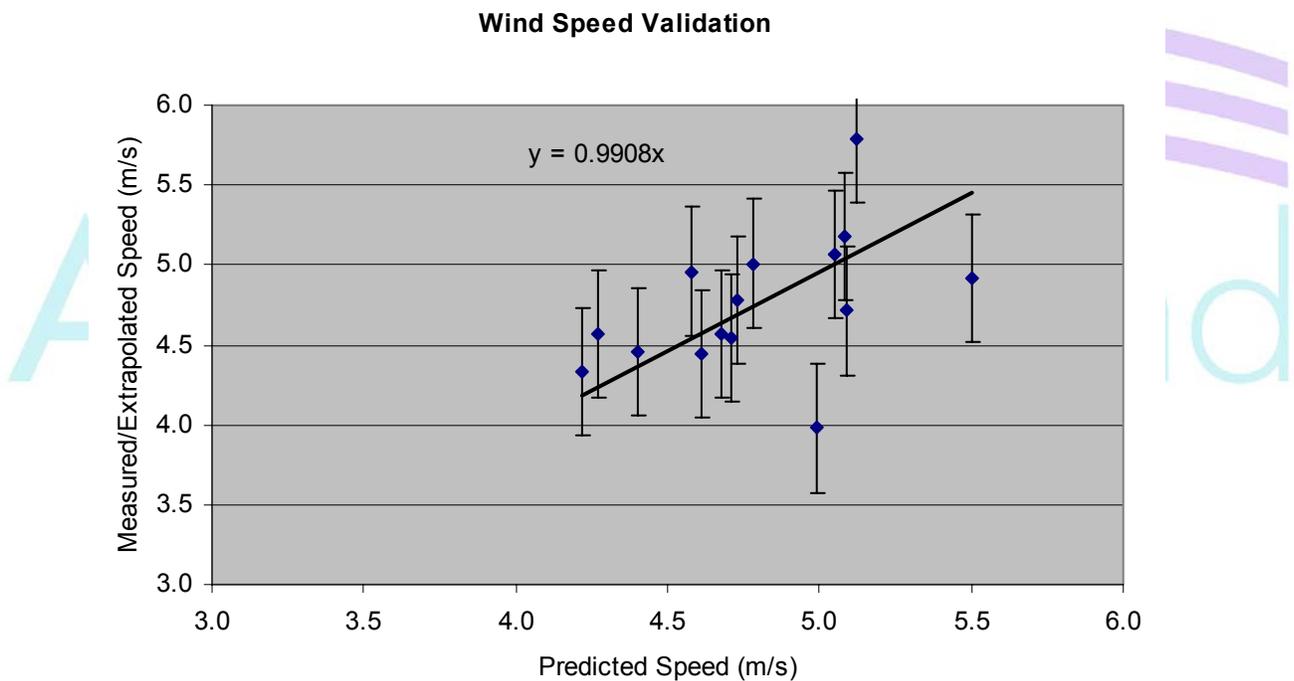


Figure 1. Comparison of the speed map with observed wind speeds at 15 airport stations projected to 50 m height. The error bars reflect period of record, tower height, and anemometer sensitivity, as described in the text. The trend line is forced through the origin.

The agreement between model and data overall is reasonably good considering the uncertainties in the extrapolated data. Although the standard deviation of the bias shown in Table 1 is, in absolute terms, within the expected range for MesoMap projects, it is somewhat larger than usual in percentage terms (the typical error rate being 5-7%). The main reason is the relatively low average speed observed at the airports.

After reviewing the validation results, AWS Truewind determined that no adjustments were necessary to the wind speed or power density maps.

5. WIND MAPS

The accompanying maps show the predicted mean annual wind speed in South Carolina at heights of 30, 50, 70, and 100 m; maps of mean annual wind power at 50 and 100 m are also provided.

The wind maps show that the best wind resource within state borders is generally found along the coastal areas and inland water bodies. The wind speed at 70 m height (a typical height for modern wind turbines) in these regions is predicted to be in the range of 5.5 to 7.0 m/s with an average between 6.0 and 6.5 m/s. The wind power is predicted to range from 100 to 400 W/m², or NREL class 1 through 3, with a mean power density ranging from 200 to 300 W/m². The remainder of the state has predicted wind speeds less than 5.5 m/s with some locations being as low as 2.5 m/s. The wind power density in these areas averages from 100 to 200 W/m², NREL class 1- and 1+ with the lowest wind power density being 15 W/m².

The main reasons for the generally low wind speeds onshore are the high surface roughness caused by trees; relatively flat terrain throughout most of the state; and South Carolina's position south of jet stream and the normal winter storm track across North America. However, as the elevation increases further inland and reaches over 1000m, a few ridgelines can be found along the boundary of North Carolina that are predicted to experience wind speeds of up to 8.0 m/s.

The predicted wind speeds are higher offshore, ranging from 6.5 to 7.5 m/s near the coast to 8.0 and 8.5 m/s farther offshore. Here, wind power density ranges from 300 to 400 W/m² nearest to the shore and 600 to 800 W/m² further offshore. This coincides with an NREL wind resource class from 3 to 6.

It should be stressed that the mean wind speed at any particular location may depart substantially from the predicted values, especially where the elevation, exposure, or surface roughness differs from that assumed by the model, or where the model scale is inadequate to resolve significant features of the terrain.

6. GUIDELINES FOR USE OF THE MAPS

The following are guidelines for interpreting and adjusting the wind speed estimates in the maps, to be used in conjunction with the accompanying ArcReader CD. The ArcReader CD allows users to obtain the "exact" wind speed value at any point, and it provides the elevation and surface roughness data used by the model, which are needed to apply the adjustment formulas given below.

1. The maps assume that all locations are free of obstacles that could disrupt or impede the wind flow. “Obstacle” does not apply to trees if they are common to the landscape, since their effects are already accounted for in the predicted speed. However, a large outcropping of rock or a house would pose an obstacle, as would a nearby shelterbelt of trees or a building in an otherwise open landscape. As a rule of thumb, the effect of such obstacles extends to a height of about twice the obstacle height and to a distance downwind of 10-20 times the obstacle height.
2. Generally speaking, points that lie above the average elevation within a 200×200 m grid cell will be somewhat windier than points that lie below it. A rule of thumb is that every 100 m increase in elevation will raise the mean speed by about 0.5 m/s. This formula is most applicable to small, isolated hills or ridges in flat terrain.
3. The mean wind speed at a location could be affected by the roughness of the land surface – determined mainly by vegetation cover and buildings – up to several kilometers away. If the roughness is much lower than that assumed by the model, the mean wind speed could be higher. Typical values of roughness range from 0.01 m in open, flat ground without significant trees or shrubs, to 0.1 m in land with few trees but some smaller shrubs, to 1 m or more for areas with many trees. These values are only indirectly related to the size of the vegetation.

The following equation provides an approximate speed adjustment for differences in surface roughness in the direction of the wind:

$$\frac{v_2}{v_1} \approx \frac{\log\left(\frac{300-d}{z_{01}}\right)}{\log\left(\frac{h-d}{z_{01}}\right)} \times \frac{\log\left(\frac{h-d}{z_{02}}\right)}{\log\left(\frac{300-d}{z_{02}}\right)}$$

v_1 and v_2 are the original and adjusted wind speeds at height h (in meters above ground level); z_{01} and z_{02} are the model and actual surface roughness values (in meters); and d_1 and d_2 are the corresponding displacement heights. (This equation assumes the wind is unaffected by localized roughness changes above a height of 300 m.)

As an example, suppose the surface roughness assumed by the model was 0.2 m, and the displacement 2 m, whereas the true roughness is 0.75 m and displacement 7.5 m. For $h = 50$ m, the above formula gives

$$\frac{v_2}{v_1} \approx \frac{\log\left(\frac{300-2}{0.2}\right)}{\log\left(\frac{50-2}{0.2}\right)} \times \frac{\log\left(\frac{50-7.5}{0.75}\right)}{\log\left(\frac{300-7.5}{0.75}\right)} = 0.90$$

This shows that the predicted wind speed should be reduced by about 10%.

This formula assumes that the wind is in equilibrium with the new surface roughness above the height of interest (in this case 50 m). When going from high roughness to low roughness (such as from forested to open land), the clearing should be at least 1000 m wide for the benefit of the lower roughness to be fully realized. However, when going from low to high roughness, the reduction in wind speed may be felt over a much shorter distance. For this and other reasons, the formula should be applied with care.

