

Title of my master thesis

Franziska Hellmuth



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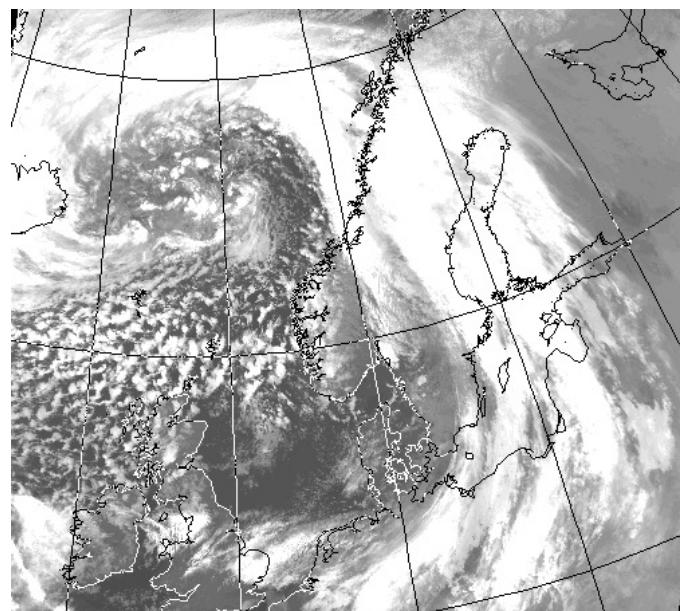
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ABSTRACT

This abstract is from the GEO4012 Literature study course. And has to be updated! Previous studies show that retrievals can lead to non-unique snowfall accumulation. Regional models are still under development to represent mesoscale features precisely. The Haukeliseter site is used to measure snow accumulation since winter 2010 / 2011. This new approach makes it unique since it is exposed to high wind speeds and snow amounts of 2 - 3 m are observed during a regular winter season. During Christmas 2016 an extreme weather event occurred at Haukeliseter. High wind speeds combined with high precipitation of snow and rain were monitored. Optimal estimation retrievals are under construction to measure the amount of snowfall accumulation. Retrievals such as the one for CloudSat try to avoid the non-uniqueness of different snowfall habits and uses a priori information such as temperature. Another approach is to use microphysical information from MASC and PIP to reduce uncertainties in the radar snowfall retrieval. The MetCoOP ensemble prediction system is operational at the Norwegian Meteorological institute since 2016. Preliminary comparison between MEPS and double fence gauge measurements show an overestimation of snowfall accumulation with longer lead time.

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LIST OF ABBREVIATIONS

| | |
|-------------------|---|
| AR | Atmospheric River |
| AROME | Applications of Research to Operations at Mesoscale |
| C3VP | Canadian CloudSat-CALIPSO Validation Project |
| CPR | Cloud Profiling Radar |
| DT | Dynamic Tropopause |
| ECMWF | European Centre for Medium-Range Weather Forecasts |
| EPS | Ensemble Prediction System |
| FMI | Finnish Meteorological Institute |
| IVT | Integrated Vapour Transport |
| MASC | Multi-Angular Snowfall Camera |
| MEPS | MetCoOp Ensemble Prediction System |
| Met-Norway | Norwegian Meteorological Institute |
| MetCoOp | Meteorological Co-operation on Operational NWP |
| MRR | Micro Rain Radar |
| MSLP | Mean Sea Level Pressure |
| NWP | Numerical Weather Prediction |

| | |
|-------------|---|
| PIP | Precipitation Imaging Package |
| PSD | Particle Size distribution |
| SMHI | Swedish Meteorological and Hydrological Institute |
| SWC | Snow Water Content |
| SWP | Snow Water Path |
| WCB | Warm Conveyor Belt |
| WMO | World Meteorological Organization |

CHAPTER 1: INTRODUCTION

The overleaf file from the Introduction is here:

<https://www.overleaf.com/13946064tcvpbpwjzjnk>

During Christmas 2016 a storm impinged on the west coast of Norway. The storm, called 'Urd' was according to [Olsen and Granerød \[2017\]](#) associated with strong winds and high precipitation amounts. The midwind along the coast of Western Norway had hurricane strength (observed: 40 m s^{-1} to 55 m s^{-1}). In South and Eastern Norway west to north-west winds between 25 m s^{-1} to 40 m s^{-1} were measured. At the Haukeliseter measurement site, 136.4 mm of precipitation were monitored during 21 December 2016 to 27 December 2016. This event was just above the limit of been called an extreme weather. Storms of this kind are expected to occur on average every five years [[Olsen and Granerød, 2017](#)].

The financial costs associated with 'Urd' are estimated to about 180 million Norwegian kroner. 'Urd' led to major traffic problems for cars, trains, ferries and air planes. Most mountain crossings were kept closed during Christmas 2016. In addition, there was a power breakout of around 70.000 households and 40 emergency power stations failed during the extreme weather.

This extreme weather, might not have lead to the same damages as some of the extreme weather events of recent years. But since people are affected by extreme weather (Figure 1.0.1) it is important to predict storms, associated precipitation and wind as accurately as possible. Having accurate observations, will lead to better performing models which rely on observations. [include a reference here](#)

Figure 1.0.1 shows that precipitation and strong winds can influence in certain ways the infrastructure. To predict and measure snowfall accumulation as accurately as possible is important since snowfall has impact on avalanches, freshwater release into water systems

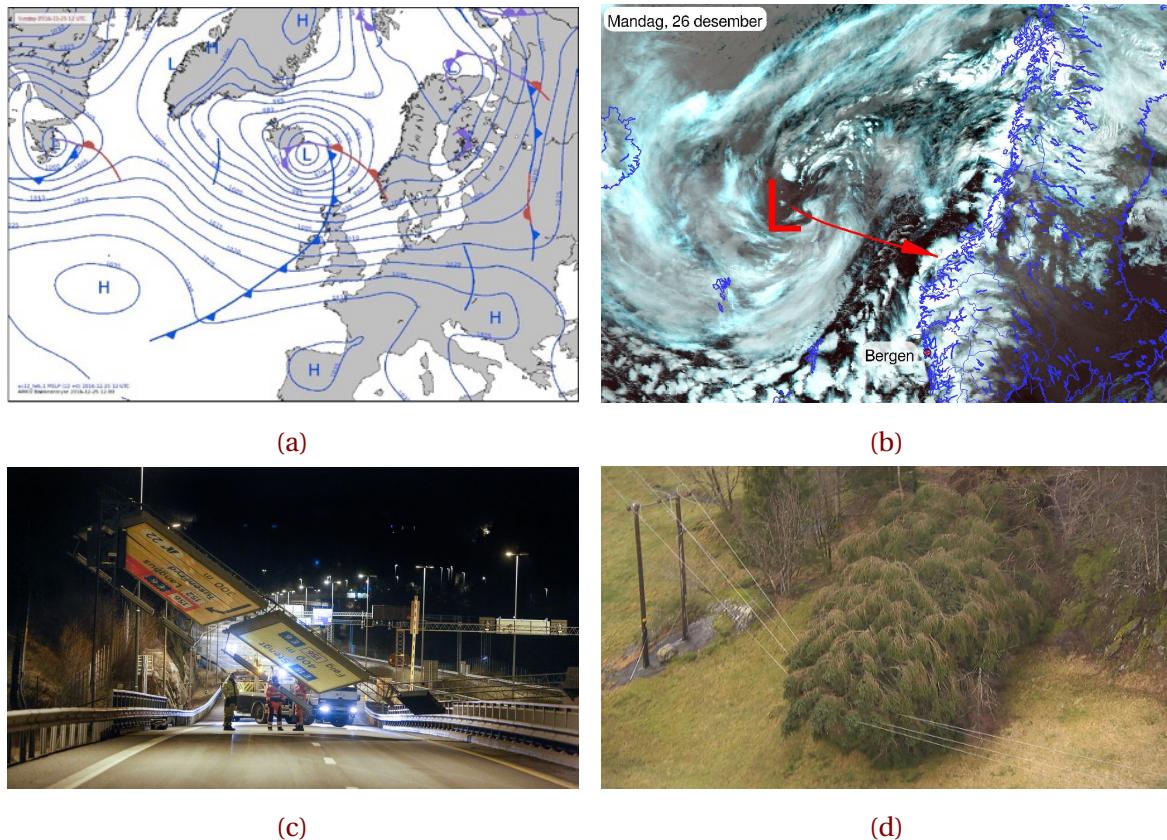


Figure 1.0.1: In a: Weather situation Sunday 25 December 2016 at 12 UTC, [Olsen and Granerød, 2017]. b: Tweet from Meteorologene [2016]: Here comes #Urd! The low pressure centre will hit Møre og Romsdal, but the strongest wind comes south of Stad. #SørNorge. c: This traffic sign, ten meter long and four meter high was blown down during the storm, [Ruud et al., 2016]. d: Trouble maker: The extreme weather during Christmas created problems for the local infrastructure. 80.000 households were without electricity during the storm, [Farestveit, 2016].

in spring, and extra economical expenses for local infrastructure as well as climatological effects.

Joos and Wernli [2012] investigated the influence of microphysical processes on potential vorticity development in warm conveyor belts (WCB). They demonstrated the complex interaction between the small-scale microphyiscal processes and the large-scale flow in WCB. For the understanding of numerical simulations of storm developments it is import-

ant to know vertical precipitation profiles and their position within the synoptic vorticity environment. It is therefore crucial to study the vertical structure of different synoptic storms and predict as accurately as possible.

Since November 2016, the Meteorological Cooperation on Operational Numerical Weather Prediction (MetCoOP) Ensemble Prediction forecast (MEPS) is operational at the Meteorological Institute of Norway (Met-Norway). The study by Müller et al. [2017] shows that the AROME-MetCoOp, a version of the Météo-France Applications of Research to Operations at Mesoscale, performs well for certain meteorological phenomena.

Microphysical processes in weather models are still not well understood and therefore are mostly parametrised [Müller et al., 2017]. Furthermore, high latitude regions are not well represented in meteorological models. Indeed a comparison between the MEPS data fit the observations for December 2016 but uncertainties are still present for this time period. Some satellites, such as CloudSat have been equipped with radar to estimate snowfall rates and vertical profiles of precipitation. CloudSat is one of the satellites orbiting in the A-Train formation and measures the vertical structure of cloud systems [Stephens et al., 2002].

Studies of Kulie and Bennartz [2009] showed that the Cloud Profiling Radar (CPR), mounted on the CloudSat, can be used to estimate global distributions of snowfall. They showed that different combinations of microphyiscal habits and fall speed can lead to the same results of reflectivity and therefore to the same amount of snowfall rate. Methods like optimal estimation retrieval were established to reduce the non-uniqueness. Where ground observations are used to estimate vertical profiles of precipitation.

The improvement of the CloudSat retrieval is helpful to show that climate models over estimate present-day Antarctic snowfall [Palerme et al., 2017]. Norin et al. [2015] presented a good agreement between the ground-based snowfall measurements and satellite observations.

1.1 MEASUREMENT SITE - HAUKELEISETER

Haukeliseter, shown in Figure 1.1.1 is a mountain plateau 991 m above sea level, located in the Norwegian county 'Telemark' (59.8° N, 7.2° E). The station measures precipitation, temperature, snow depth and wind. It has served as a measurement site for snow accumu-

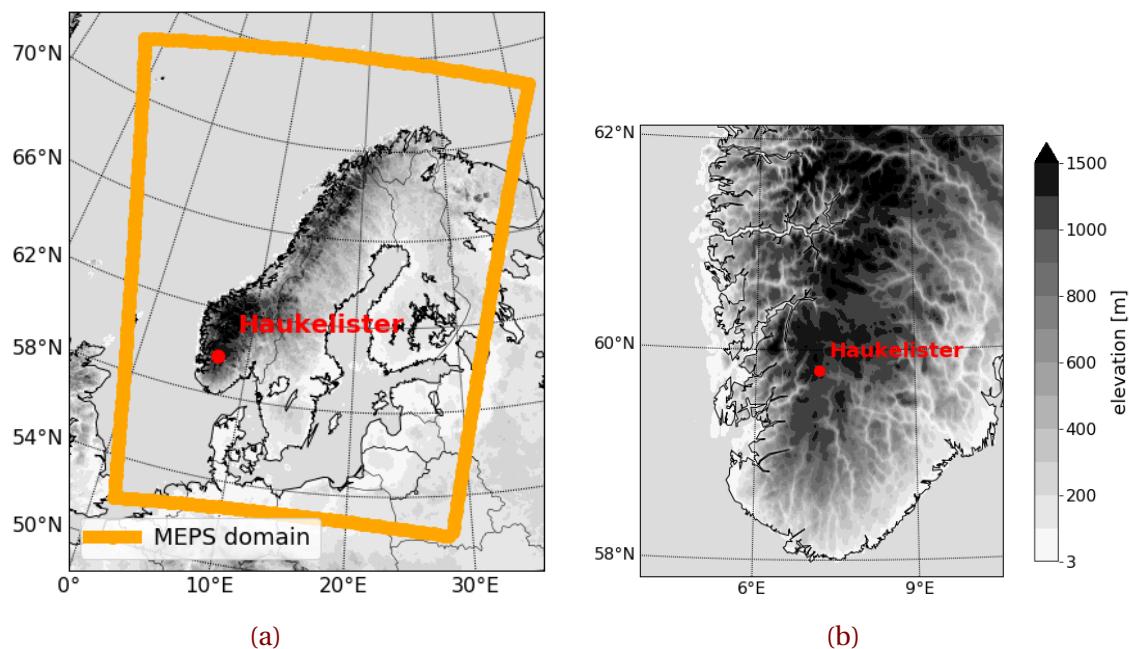


Figure 1.1.1: Elevation map of Northern Europe and South Norway, **a, b** respectively. Red dot indicates the location Haukeliseter and the orange square in **a** indicates the model domain of MEPS. Elevation according to the shading.

lation since the winter of 2010/2011 [Wolff et al., 2010, 2013, 2015].

In a study by Wolff et al. [2015] the wind-induced under-catch of solid precipitation is determined. Dependent on the kind of precipitation the wind plays different roles in the amount of accumulation. For temperatures below -2°C the wind speed influences the falling snow. Where less precipitation can be observed at higher wind speeds or more precipitation can be measured if too much is blown into the gauge. The catch ratio between the standard Geonor precipitation gauge and the DF-Geonor shows, that only 80 % of solid precipitation are observed at wind speeds of 2 m s^{-1} and only 40 % at 5 m s^{-1} , [Figure 5 in Wolff et al., 2015]. The double fence gauge is more accurate than the single fence and is used as the reference gauge. Nevertheless, this shows the need of a combination of ground based observations together with an optimal estimation retrieval to verify the accuracy of MEPS. Wolff et al. [2015] introduced an adjustment function for the Geonor double fence, so that different precipitation under certain wind speeds are presented correctly and can be used as confidential data.

1.2 OBSERVATIONS IN DECEMBER

The general climate at Haukeliseter can be defined with the updated Köppen-Geiger climate types presented in [Peel et al. \[2007\]](#). Figure 8 in [Peel et al. \[2007\]](#) shows, that Haukeliseter may lay in a transition zone and can be categorized as ET, a polar tundra climate type (hottest month temperature $T_{hot} \geq 0^{\circ}\text{C}$) or Dfc, a cold climate without dry season and cold summers.

Haukeliseter presents a typical Norwegian climate condition. At the measurement site, frequent snow events combined with high wind speeds are observed during a six to seven month winter period. In addition a snow amount of about 2 m to 3 m can be expected, where 50 % of the yearly precipitation is solid [[Wolff et al., 2010, 2013, 2015](#)].

The mean wind direction for solid precipitation is from the west with maximum wind speeds above 15 m s^{-1} , observed during a 10 year winter period at a nearby station [[Wolff et al., 2010, 2015](#)]. As seen in Figure 1.2.1 is the average December temperature -6°C (30-yr period 1961 to 1990, value taken from [eklima \[2016\]](#)). December 2016 was warmer with an anomaly of +4.9 K above the climate mean. In 2016, the precipitation was 200 % more than the climate mean. [yr.no says something different. According to them was it only 76 %](#) The precipitation observed in the time period 21 December 2016 to 27 December 2016 where 56.9 % of the total accumulation in December 2016. Furthermore, a maximum wind of 22.3 m s^{-1} was observed in this period, which can be associated to a slight storm.

1.3 PRELIMINARY MEPS AND OBSERVATION RESULTS

As MEPS is operational since November 2016 one can compare the ensemble forecast model output with observations from the double fence at Haukeliseter. This will later on be compared to the vertical optimal snowfall retrieval estimates.

During Christmas 2016 a storm approached the Haukeliseter site, which resulted in precipitation in form of liquid and solid. Figure 1.3.1 shows the preliminary comparison between the MEPS ensemble member forecasts as well as the snowfall accumulation measured by the double fence at Haukeliseter. The double fence data is noisy and therefore, a filtered dataset is accessed from [eklima \[2016\]](#).

Each MEPS cycle is initialized at 18 UTC for the respective day. Grey lines in Figure 1.3.1

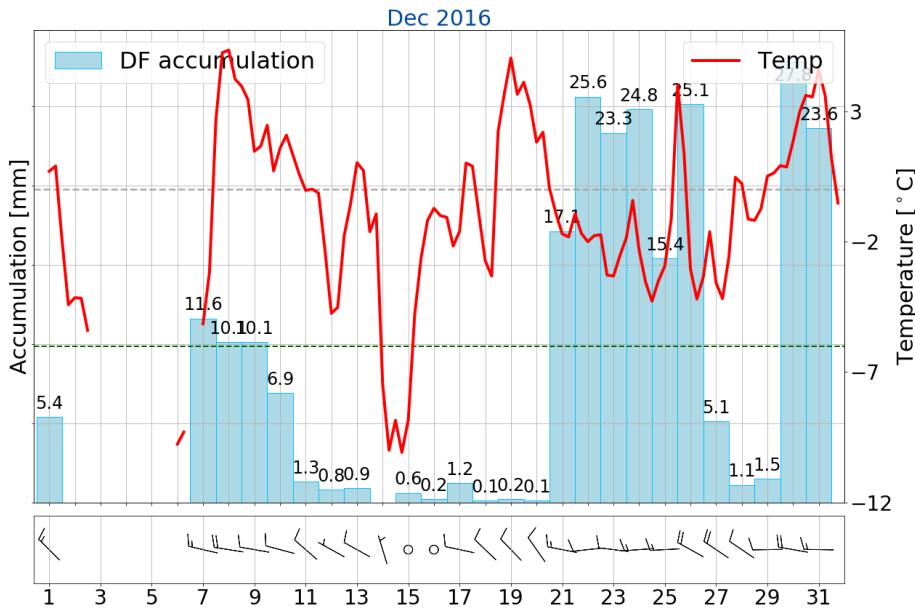


Figure 1.2.1: Observations at Haukeliseter weather mast for December 2016. Daily accumulation [mm] in light blue, mean temperature every six hours (red, [°C]), and daily maximum wind as barbs [m s^{-1}]. Gray dashed line indicates the freezing temperature. The monthly normal value is green dashed ($-6.0\text{ }^{\circ}\text{C}$), the values are taken from [eklima \[2016\]](#). Note, that no data was available from 2 December 2016 and 6 December 2016

show the nine perturbed ensemble members of MEPS. Where the black line reflects the control run, and the dots accumulation by the double fence.

During the first few days the ensemble outputs cover the amount of snow good in comparison to the double fence observations. The spread of the ensemble members around the control run fits as well to the observations for this time period. But, for an initialisation on the 24 December 2016, 18 UTC one can see that MEPS over estimates the amount of snow accumulation. It is even more pronounced with the initialisation on the 25 December 2016, 18 UTC (compare Figure 1.3.1f).

This can have different reasons. One of them can be that the large scale weather situation was more predictable for the first four days (Figures 1.3.1a to 1.3.1d). According to Müller et al. [2017] are strong precipitation events better predicted with MEPS than ECMWF (European Centre for Medium-Range Weather Forecasts). On the other side, Müller et al. [2017] states, that an overestimation appears, where the precipitation event (12 h accu-

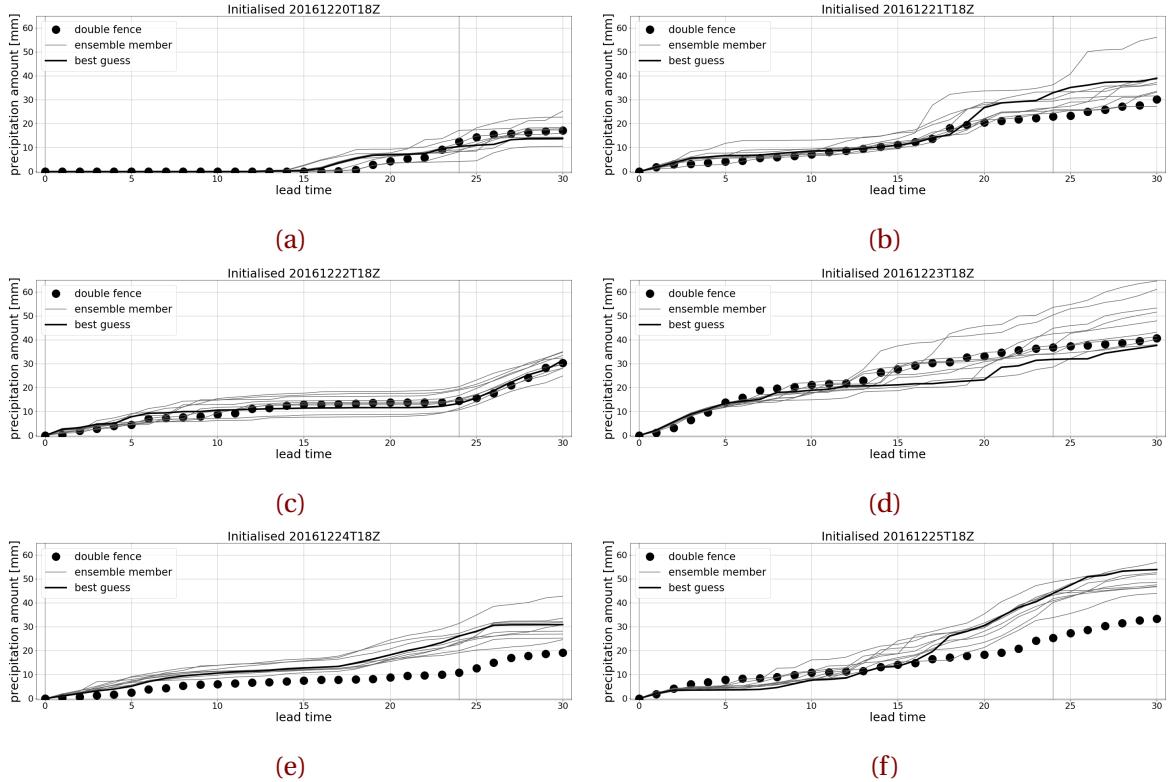


Figure 1.3.1: Accumulation of precipitation at Haukeliseter. Initialisation of MEPS at 18 UTC. Ensemble member as line in grey and the control in black. Dots indicate the hourly accumulation observed at the double fence, [eklima, 2016]. **Do you think I have to make something bigger???** Font size etc? Or use one page only for the figures and have them underneath each other?

mulation) is less than 10 mm. This is not the case for the time period 21 December 2016 to 26 December 2016 (compare Figure 1.2.1), where the daily precipitation exceed 10 mm every day. The question arises why MEPS covers the high amount of snow accumulation on the 24 December 2016 but forecasts 24 h prior to 25 December 2016 and 26 December 2016 are not covering the predicted amount of accumulation!

CHAPTER 2: INSTRUMENTS

The overleaf file from the Methodology is here:

<https://www.overleaf.com/13946091vphmgxbxpyg>

Many factors such as humidity and temperature contribute to snowflake geometry. The knowledge of snowflake habits, particle size distributions, and fall speed lead to a reduction of errors in optimal estimation retrievals.

This work is based on several datasets collected at the Haukeliseter measurement site, 59.8° N, 7.2° E. A composition of advanced ground based observations and the CloudSat precipitation retrieval will help to get a better understanding of the vertical structure of the atmosphere.

A collaboration between the University of Utah, University of Wisconsin and Met-Norway made it possible to install three additional instruments at the measurement site during winter 2016/2017. A Multi-Angle Snowflake Camera (MASC) and a Precipitation Imaging Package (PiP) will be used to determine the snow habit, the snowfall particle size distribution, and near-surface fall speed. Additionally, a Micro Rain Radar (MRR) is established to obtain fall speed and particle reflectivity aloft. Together with temperature observations at the surface, is this a good basis to reduce the non-uniqueness of snow accumulation in optimal estimation snowfall retrieval, described in Chapter 3.

2.1 DOUBLE FENCE

Since the winter season 2010/2011 Haukeliseter is equipped with three rain gauges. The wind shielded gauges are placed perpendicular to the main wind direction (Figure 2.1.1). The precipitation gauge protected by an octagonal double fence will be the reference to all surface accumulation measurements. The double fence creates an artificial calm wind and maximize the catch of precipitation, [Wolff et al., 2010, 2013, 2015]. This will get some more description. I have to read some up



Figure 2.1.1: Double fence and unprotected precipitation gauges at Haukeliseter, from Wolff et al. [2015]. The prevailing wind from east comes from the left side in the image.

2.2 MRR - MICRO RAIN RADAR

Radars are very useful to observe the vertical of the atmosphere. The instrument is able to detect mesoscale features and makes it possible to see the vertical structure of storms [Markowski and Richardson, 2011].

The principle of radar measurements is based on an electromagnetic wave, which is emitted from the radar transmitter and interacts with the hydrometeors along the beam. A fraction of the pulse energy is reflected back to the receiver of the radar. The quantity of scattering depends on the shape and structure of the reflected particle. This creates vertical profiles of reflectivity. The reflectivity gives information about the diameter of the object.

The Micro Rain Radar, in Figure 2.2.1, measures profiles of Doppler spectra [METEK, 2010]. The Doppler spectrum tells about the movement of the particle. The vertical pointing Doppler radar measures the energy that is returned from each interval and thus enabling the detection of the Doppler spectrum [L'Ecuyer, 2017]. The MRR measures



Figure 2.2.1: Micro Rain Radar.

at a frequency of 24 GHz, the temporal and spatial resolution of 60 s and 100 m, respectively. The height ranges from 100 m (because of ground clutter) to 3.000 m [METEK, 2010].

Table 2.2.1: Typical reflectivity values, from Doviak and Zrnic [1993]. The values are obtained from measurements, models and observations. The rainfall rate R is calculated with Equation (2.2.2).

| | Ze [dBZ] | R [mmh ⁻¹] |
|----------------------------|--------------------|----------------------------------|
| Drizzle | <25 | 1.3 |
| Rain | 25 to 60 | 1.3 to 205.0 |
| Snow | | |
| dry, low density | <35 | 5.6 |
| Crystal; dry, high density | <25 | 1.3 |
| wet, melting | <45 | 23.7 |
| Graupel | | |
| dry | 40 to 50 | 11.5 to 48.6 |
| wet | 40 to 55 | 11.5 to 99.9 |
| Hail | | |
| small; <2 cm, wet | 50 to 60 | 48.6 to 205.0 |
| large; >2 cm, wet | 55 to 70 | 99.9 to 864.7 |
| Rain & Hail | 50 to 70 | 48.6 to 864.7 |

Radar reflectivity (z) from the MRR is transformed from $1 \text{ mm}^6/\text{m}^3$ to dBZ. The transformations is done by the following relationship;

$$Ze = 10 \log_{10} \left(\frac{Z}{1 \text{ mm}^6/\text{m}^3} \right) \quad [\text{dBZ}] \quad (2.2.1)$$

A transformation to rainfall rates can be performed by the Z - R relationship. By the use of typical fall speeds and the Marshall and Palmer particle size distribution for liquid

particles, the rainfall rate in each layer can be estimated [Rinehart, 2010].

$$Z = 200R^{\frac{8}{5}} \quad [\text{mm}^6 \text{m}^{-3}]$$

$$R = \left(\frac{10^{\frac{Ze}{10}}}{200} \right)^{\frac{8}{5}} \quad [\text{mm h}^{-1}] \quad (2.2.2)$$

Table 2.2.1 represents the Z-R relationship if the Marshall-Palmer assumption (Equation (2.2.2)) is applied. Z-snowfall relationships are developed but are difficult to apply due to the variation of size and density of the particles.

After the transformation to dBZ the reflectivity is averaged every 200 m layer. For this, all values above 300 m are taken. A reflectivity represented at 400 m represents the average reflectivity between 300 m and 500 m.

2.3 PIP - PRECIPITATION IMAGING PACKAGE

The precipitation imaging package (PiP) is a modification of the Snowflake Video Imager presented by Newman et al. [2009]. The video distrometer is a construct of a halogen flood lamp and a video system (Figure 2.3.1). The instrument determines the habit of snowflakes from images at a frequency of 60 Hz. Lamp and lens have a distance of approximately 3 m which follows a field of view: 32 mm by 24 mm. Hence snowflakes from grey-scale images, particle size distribution (PSD) and fall-speed of precipitation can be determined.

In front of the halogen lamp is a frosted window, so that the background light is uniform over all time. A falling particle appears as a 2-D shadow in the video image. Newman et al. [2009] describes in detail the algorithm applied to the system to get information about the snow-particle habit.



Figure 2.3.1: Precipitation Imaging Package.

Winds have almost no effect on the result of the video distrometer [Newman et al., 2009]. They also say, to reduce eventual wind effects, should the distrometer be oriented with regard to storm winds (optical axis perpendicular to mean wind). Was that the case for Haukeliseter (I'm just curious)???

2.4 MASC - MULTI-ANGULAR SNOWFALL CAMERA

Instruments like the afore mentioned PiP has according to Garrett et al. [2012] coarser resolution and the determination of particle size can have size errors. Hence, a new instrument was developed. The Multi-Angular Snowfall Camera (MASC) takes high-resolution images of hydrometeors in free fall and measures the fall-speed simultaneously .

The MASC consists of three cameras, three flashes, and two near-infrared sensors, pointing at a ring centre (Figure 2.4.1). A hydrometeor has to pass through the ring in a certain way to trigger the near-infrared sensors. At the same time the three cameras take a picture of the falling particle. Since the cameras take pictures from three different angles, the particles size, shape, and orientation can be specified from the image and an algorithm, described in Garrett et al. [2012]. Furthermore, the heritage of the hydrometeor, such as collision-coalescence, riming , capture nucleation, or aggregation, can be determined as well as the form.

The near-infrared sensor, that is used to trigger the cameras and the lights quantifies the fall-speed of the hydrometeors, by measuring the time the particle needs to pass the distance between the upper and lower trigger.

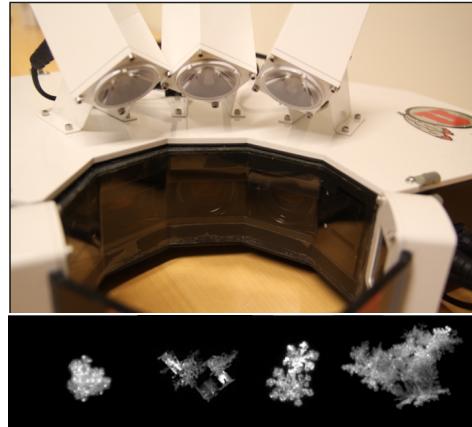


Figure 2.4.1: MASC and images taken by instrument. Lower panel taken from Cooper et al. [2017] maybe we get one for Haukeli?

CHAPTER 3: OPTIMAL ESTIMATION RETRIEVAL ALGORITHM

Since 2006, with the launch of CloudSats Cloud Profiling Radar (CPR) a global estimation of snowfall can be done. Several studies, such as [Kulie and Bennartz \[2009\]](#) have shown that estimated snowfall values depend heavily upon assumed snowflake microphysical properties. [Wood et al. \[2015\]](#) showed that a refinement of the CloudSat snowfall retrieval algorithm can be done by using snowflake models. This study was based on data from the Canadian CloudSat-CALIPSO Validation Project [C3VP, [Hudak et al., 2006](#)], where they concentrated on cold season clouds and precipitation.

In an attempt to reduce the non-uniqueness of the problem, [Wood et al. \[2015\]](#) used the a priori knowledge of snowfall microphysics and temperature (from ground-based observations) to refine the forward-model assumptions for the CloudSat snowfall retrieval scheme. Results from this scheme showed a good agreement with reported values observed at meteorological measurement sites.

Model estimates have proven, how useful the estimation retrieval can be to verify ground-based radar snowfall measurements [[Norin et al., 2015](#)]. Although the retrieval has obviously been improved the estimation algorithm, can still lead to uncertainties in the retrievals of up to 140 % to 200 % [[Wood, 2011](#)].

[Cooper et al. \[2017\]](#) developed a technique to combine MRR, MASC, and PiP information into a common retrieval framework. Specifically, estimates of snowflake microphysical properties from the MRR are used as the a priori term in the optimal-estimation retrieval scheme. The usage of either MASC / PiP or MRR fall-speed can show which a priori guess in the retrieval gives the more accurate retrieved snowfall rate at the ground.

The difference between the retrieval and the snow gauge observations was –18 % when

applied to data from Barrow, Alaska.

Cooper et al. [2017] also showed that the retrieval is sensitive to habit and fall speed. The installation of a MRR, MASC, and PIP should help to adjust the particle models for graupels and rimed particles which are often observed at Haukeliseter.

3.1 PRESENT SNOW

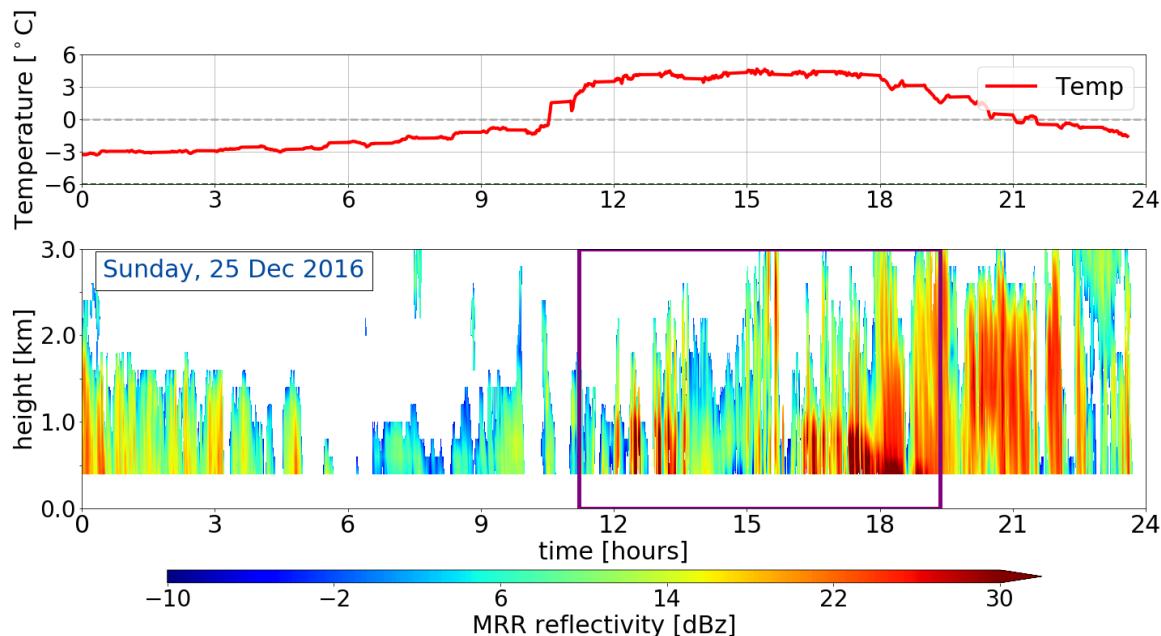


Figure 3.1.1: Shows the a priori temperature dependence within the optimal estimation retrieval for an all day precipitation event on 25 December 2016. The upper panel shows T_{ap} measured at the Haukeliseter site. The lower panel presents the reflectivity from the MRR in addition indicates the purple frame the time, where the MRR reflectivity was larger than -10 dBZ and surface temperatures less than 2°C

To achieve vertical profiles of snowfall from MRR different steps and assumptions are done in the here presented snowfall retrieval. From one of the lower levels, the snowfall rate at the surface can be estimated. The retrieval is only performed for profiles, which are likely to have observed snow, where most retrievals use rain. In previous studies relationships between reflectivity and snowfall have been developed. Even if the distribution is known,

different crystal shapes led to different results. Snow densities vary significantly from storm to storm, where small particles are still Rayleigh scattered, and larger particles non-Rayleigh scattered [Gunn and East, 1954].

To obtain the likelihood of present snow a reflectivity threshold of -15 dBZ is used. This threshold is similar to the one used in Wood et al. [2013b], where it states that light liquid precipitation is related to -10 dBZ citation here. Wood [2011] compared the reflectivities in the lowest bin and adjacent bin and found, that the reflectivity above -15 dBZ are not influenced by ground clutter.

The Haukeliseter measurement site is equipped with a weather mast, measuring the air temperature every minute at two meter height (compare Figure 3.1.1, upper panel). Since the MRR measures above 300 m and only temperature measurements at the surface exists, a priori temperature (T_a) is assumed to be similar to the observed near-surface air temperature. Using a moist adiabatic lapse rate of $dT/dz = 5 \text{ K km}^{-1}$ gives T_a in each layer. Temperature measurements above 2°C read up reference Liu (2008), Wood et al. [2013b] are neglected and assumed to be present liquid precipitation.

The purple line in Figure 3.1.1 represents the time frame during 25 December 2016, where the MRR measures reflectivity is less than -15 dBZ , and a priori temperature passes the 2°C limit at the surface.

3.2 SIZE DISTRIBUTION

To determine the snowfall rate at the surface an exponential particle size distribution (PSD) is used.

$$N(D) = N_0 \exp(-\lambda D) \quad [\text{m}^{-3} \text{ mm}^{-1}] \quad (3.2.1)$$

where λ represents the PSD slope parameter and N_0 the number density. D is the particle maximum dimension evaluated from the 2D-scattering model for branched 6-arm spatial particles with porosities for reflectivity measurements at 24 GHz (see Appendix A.1).

Wood [2011] showed a linear fit between $\log(\lambda)$ and the a priori temperature, respectively $\log(N_0)$ and the a priori temperature. Since T_a varies with a moist adiabatic lapse in each layer bin the slope parameter and the number density in Equation (3.2.1) are changing

too. Using T_a in °C for each layer bin the following logarithmic assumption is used,

$$\log(\lambda) = -0.03053 \cdot T_{ap} - 0.08258 \quad [\log(\text{mm}^{-1})] \quad (3.2.2)$$

$$\log(N_0) = -0.07193 \cdot T_{ap} + 2.665 \quad [\log(\text{m}^{-3} \text{ mm}^{-1})] \quad (3.2.3)$$

Since the results from C3VP were similar to other observations and N_0 ranges over several orders of magnitude as well as λ was for the snow events non-Gaussian, for both the log-transformed values are created to achieve the state vector \mathbf{x} of unknown micropyhsical properties.

$$\mathbf{x} = \begin{bmatrix} \log(\lambda)_0 \\ \vdots \\ \log(\lambda)_{\text{nlayer}} \\ \log(N_0)_0 \\ \vdots \\ \log(N_0)_{\text{nlayer}} \end{bmatrix} \quad \text{nlayer} = 14 \quad (3.2.4)$$

3.3 SNOWFALL RETRIEVAL SCHEME

The optimal estimation method is based on Gaussian statistics. Minimizing the scalar cost function, Φ for the snowfall properties, \mathbf{x} . The cost function weights the difference between the observed reflectivity and the simulated measurements as well as the difference between the estimated and a priori guess. A forward model $F(\mathbf{x})$ relates unknown snowfall parameters \mathbf{x} to radar observations y and approximates the true physical state between them [Cooper et al., 2017, Wood et al., 2014].

Scalar cost function:

$$\begin{aligned} \Phi(\mathbf{x}, y, a) = & (y - F(\mathbf{x}))^T \mathbf{S}_y^{-1} (y - F(\mathbf{x})) \\ & + (\mathbf{x} - a)^T \mathbf{S}_a^{-1} (\mathbf{x} - a) \end{aligned} \quad (3.3.1)$$

where, \mathbf{x} , vector of retrieved snowfall properties (Equation (3.2.4)); y , vector of observation (MRR reflectivity); a , vector of the a priori guess (temperature dependent); F , forward model (Equation (3.4.6)); \mathbf{S}_a , a priori error covariance matrix; \mathbf{S}_y , measurement error covariance matrix.

\mathbf{S}_a links the uncertainties of the PSD information and the surface temperature differences. The diagonal matrix elements in \mathbf{S}_a are equal to 0.133 and 0.95 for the particle slope parameter and the number density, respectively as in Eq. 7.35 and 7.36 in Wood [2011]. \mathbf{S}_y characterises the the uncertainties associated with the measurements and the error in the forward model. This study uses for the diagonal matrix elements 2.5² UNIT! based on the study from CITATION. BECAUSE.

I don't understand the next steps and if it is still the same \mathbf{x}^* !

At convergence is the error covariance of the retrieved state vector \mathbf{S}_x

$$\mathbf{S}_x = \left(\mathbf{S}_a^{-1} + \mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1} \quad (3.3.2)$$

which follows for \mathbf{x}

$$\mathbf{x} = \underbrace{\left(\mathbf{S}_a^{-1} + \mathbf{K}^T \mathbf{S}_y^{-1} \mathbf{K} \right)^{-1}}_{\mathbf{S}_x} \left(\mathbf{S}_a^{-1} \mathbf{a} + \mathbf{K}^T \mathbf{S}_y^{-1} (y - F(\mathbf{x}) + \mathbf{K}\mathbf{x}) \right) \quad (3.3.3)$$

The Jacobian matrix, \mathbf{K} , represents the sensitivity matrix of the perturbed result of the forward model. The true state \mathbf{x} is perturbed by $\pm 0.2\%$ and thus \mathbf{K} represents the relation between simulated values to the true state. Why are we perturbing it?

$$\mathbf{K}(\mathbf{x}) = \frac{\partial \mathbf{y}}{\partial \mathbf{x}} = \begin{bmatrix} \frac{\partial y_0}{\partial \mathbf{x}_0} & \frac{\partial y_0}{\partial \mathbf{x}_1} & \cdots & \frac{\partial y_0}{\partial \mathbf{x}_{2 \times \text{nlayer}}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial y_{\text{nlayer}}}{\partial \mathbf{x}_0} & \frac{\partial y_{\text{nlayer}}}{\partial \mathbf{x}_1} & \cdots & \frac{\partial y_{\text{nlayer}}}{\partial \mathbf{x}_{2 \times \text{nlayer}}} \end{bmatrix} \quad (3.3.4)$$

Usually, is \mathbf{K} not diagonal [Wood, 2011], hence it is a mix of the true state and the a priori guess and the influence from them can be estimated. The closer \mathbf{K} is diagonal, the more is \mathbf{x} determined by the real observed and a priori values. If the limit of the partial derivative is close to unity, the retrieved value \mathbf{x} is its true state [Wood, 2011].

mmmh? What exactly are we doing here? Test the if convergent:

$$\hat{\mathbf{x}} = (\mathbf{x} - F(\mathbf{x}))^T \mathbf{S}_x^{-1} (\mathbf{x} - F(\mathbf{x})) \quad (3.3.5)$$

only if $\hat{\mathbf{x}}$ is smaller than 2 it is a 'good' retrieval.

To test the result of \mathbf{x} a χ^2 test is performed at the convergence of \mathbf{S}_x .

$$\begin{aligned} \chi^2 = & (F(\mathbf{x}) - y)^T \mathbf{S}_y^{-1} (F(\mathbf{x}) - y) \\ & + (\mathbf{x} - \mathbf{a})^T \mathbf{S}_a^{-1} (\mathbf{x} - \mathbf{a}). \end{aligned} \quad (3.3.6)$$

The first term in Equation (5.3.1) measures the part of χ^2 related to the noise of the forward model, and the second part the relation to the state vector. Thus the second term describes the accuracy of the quantities within the reflectivity and temperature measurements [Rodgers, 2000]. Furthermore, are the error contribution from the reflectivity measurement uncertainty, \mathbf{S}_{y_e} , and the uncertainty of the a priori values, \mathbf{S}_{a_e} estimated. **Where comes this formulas from? Is the \mathbf{D}_y the \mathbf{K}_b in Eq. 15 in Wood et al. [2013a]? How to interpret?**

$$\mathbf{D}_y = \mathbf{S}_x \mathbf{K}^T \mathbf{S}_y^{-1} \quad \mathbf{S}_{y_e} = \mathbf{D}_y \mathbf{S}_y \mathbf{D}_y^T \quad (3.3.7)$$

$$\mathbf{D}_a = \mathbf{S}_x \mathbf{S}_a^{-1} \quad \mathbf{S}_{a_e} = \mathbf{D}_a \mathbf{S}_a \quad (3.3.8)$$

The diagonal of \mathbf{S}_{y_e} or \mathbf{S}_{a_e} is a first estimate of the retrieval noise related to the observations [Rodgers, 2000]. Uncertainties in \mathbf{S}_{a_e} occur because of the variability in T_a and similar for \mathbf{S}_{y_e} due to measurement errors from the MRR.

To achieve snowfall rates at the surface, the snow water content (SWC) has to be transformed into snowfall rates. The use of an assumed particle fall speed of $V = 0.85 \text{ m s}^{-1}$ and the retrieved SWC (Equation (3.4.1)) gives the snow mass flux $J_{snow} = \text{SWC} \cdot V$ in $[\text{kg m}^{-2} \text{ s}^{-1}]$. **Why did we use this fall speed? Where does this assumption come from? Similar as Cooper et al. [2017] Eq. 4?** To compare retrieved snow fall rates to double fence measurements and the forecast model output, the precipitation amount at the surface is calculated.

$$P = J_{snow} \times 10^{-3} \cdot (3600 \text{ s} \cdot 24) \quad [\text{mm d}^{-1}] \quad (3.3.9)$$

The precipitation amount at the surface, presented in Chapter 7, are taken to be equal to the values at the snow layer in 800 m. The use of the values at 800 m is due to the small increase of reflectivity (ground clutter) in the bottom layers and would follow more observed snow.

- **Not sure what to do with this!**
- The error from the retrieved state vector \mathbf{x} is calculated

$$\pm \mathbf{S}_{x_{err}} = \mathbf{x} \pm \mathbf{S}_x$$

$$equiv_{err} = \frac{1}{2} \left(\frac{|\text{SWC}(-\mathbf{S}_{x_{err}}) - \text{SWC}(+\mathbf{S}_{x_{err}})|}{\text{SWC}(-\mathbf{S}_{x_{err}})} + \frac{|\text{SWC}(-\mathbf{S}_{x_{err}}) - \text{SWC}(-\mathbf{S}_{x_{err}})|}{\text{SWC}(-\mathbf{S}_{x_{err}})} \right)$$

3.4 FORWARD MODEL

Forward model defines a relationship between the radar observations and the retrieved state vector \mathbf{x} . The knowledge about the a priori parameters and related covariances, as well as \mathbf{x} , are used to minimize Equation (3.3.1). The values of \mathbf{x} are found by Newtonian iteration [Wood et al., 2014, Eq. 5].

The snow water content in each layer is estimated from the knowledge of the snow particle mass-dimension relationship in Appendix A.1, and a given slope parameter and number density (Equation (3.2.4)).

$$\text{SWC} = \int_{D_{min}}^{D_{max}} m(D) N(D) dD \quad [\text{gm}^{-3}] \quad (3.4.1)$$

To achieve a relationship between the reflectivity and the snowfall amount one needs to account for attenuation in the atmosphere. Using the previously calculated PSD (Equation (3.2.1)) the singly-scattered non-attenuated reflectivity for Rayleigh approximation ($2\pi D/\lambda \ll 1$, λ : wavelength of incident radiation) is found [Kulie and Bennartz, 2009, L'Ecuyer and Stephens, 2002, Wood et al., 2015].

$$\begin{aligned} \eta_{bk} &= \int_{D_{min}}^{D_{max}} N(D) \sigma_{bk} dD \quad [\text{m}^{-1}] \\ Z e^{ss,na} &= \frac{\Lambda^4}{\|K_w\|^2 \pi^5} \eta_{bk} \quad [\text{mm}^6 \text{m}^{-3}] \end{aligned} \quad (3.4.2)$$

where, $\|K_w\|^2$ is the complex refractive index of water and varies between 0.91 and 0.93 for wavelength between 0.01 m and 0.10 m and is independent of temperature. It also exists a complex refractive index for ice $\|K_i\|^2$, which is 0.18. This is valid for a density of 0.917 gcm^{-3} and is independent of temperature and of wavelength in the microwave region [Doviak and Zrnic, 1993]. In this work $\|K_w\|^2 = 0.93$ is chosen, BECAUSE???. σ_{bk} represents the backscattering cross-section and Λ is the wavelength OF the radar.

The singly-scattered reflectivity has to be corrected for attenuation in the layers above the actual layer. According to Beer's law is in a homogeneous medium one way transmission assumed.

$$\frac{I_\lambda}{I_{\lambda_0}} = \exp \left[- \int_0^s \beta_{ext} ds' \right] \quad (3.4.3)$$

where s is the path length through the medium. The transmissivity $I_\lambda / I_{\lambda_0}$ is the relation of survived radiation through extinction in the atmosphere with the snow extinction coefficient β_{ext}

$$\beta_{ext} = \int_{D_{min}}^{D_{max}} N(D) \sigma_{ext} dD \quad [\text{m}^{-1}] \quad (3.4.4)$$

The extinction coefficient is the sum of absorption and scattering in the atmosphere followed from the extinction cross-section $\sigma_{ext} = \sigma_{abs} + \sigma_{scat}$, [Lamb and Verlinde, 2011, Lohmann et al., 2016]. [Eq. 12.1 and more Lohmann et al., 2016]

Following Equations (3.4.2–4) the singly-scattered attenuated reflectivity $Ze^{ss,a}$ is

$$Ze^{ss,a} = Ze^{ss,na} \cdot \frac{I_\lambda}{I_{\lambda_0}} \quad [\text{mm}^6 \text{m}^{-3}] \quad (3.4.5)$$

That follows for the simulated reflectivity from the forward model, $F(\mathbf{x})$ in Equation (3.3.1),

$$F(\mathbf{x}) = \begin{bmatrix} Ze_1^{ss,a} \\ \vdots \\ Ze_{nlayer}^{ss,a} \end{bmatrix} \quad [\text{dBZ}] \quad (3.4.6)$$

- not really using the following - What to do with it?
- total number concentration

$$n_{tot} = \int_{D_{min}}^{D_{max}} N(D) dD \quad [\text{m}^{-3}]$$

- ... ???

CHAPTER 4: NUMERICAL FORECAST MODEL

MEPS was newly operational at Met-Norway when the extreme weather occurred in Norway. Comparing model data with actual observations helps to verify the agreement between model prediction and ground based measurements.

AROME-MetCoOp was operational from March 2014 until November 2016, when it was replaced with an ensemble prediction system (EPS) based on AROME-MetCoOp. MEPS is used as weather forecast at the Norwegian Meteorological Institute, the Swedish Meteorological and Hydrological Institute (SMHI) and the Finnish Meteorological Institute (FMI), [Køltzow, 2017, Müller et al., 2017].

4.1 AROME - METCOOP

In principle, MEPS is a short-term weather forecast of 66 h with 10 ensemble member and a horizontal resolution of 2.5 km and 65 vertical levels. In the following section, the model configuration will be briefly explained.

The orange frame in Figure 1.1.1 shows the MEPS model domain as it was operational for December 2016. It covers the Nordic Countries including open water such as the Atlantic Ocean, the North and the Baltic Sea.

The centre of the model is approximately at 63.5° N, 15° E. The horizontal grid points are projected on a Lambert projection to receive the same area size of each grid cell. The outer, parent grid is the ECMWF-IFS model (European Centre for Medium-Range Weather Forecasts Integrated Forecasting System) with a horizontal resolution of 9 km [Homleid and Tveter, 2016].

Vertical hybrid coordinates are terrain-following and are mass-based, [Müller et al., 2017]. How the vertical hybrid coordinates were transformed into layer thickness or height is described in ?? . Furthermore, MEPS underlies non-hydrostatic dynamics, MetCoOp Wiki [2017].

The representation of snow is covered by a modification of the three-class ice parametrization (ICE3) scheme. Where liquid-phase processes are separated from slow ice-phase processes. To model the snow cover an one-layer atmosphere model scheme is implemented. This includes three variables such as: snow water equivalent (SWE), snow density, and snow albedo [Müller et al., 2017].

The MEPS forecasting system consist of 1+9 members where each of the perturbed members perform an initialization of 66 h at 0 UTC, 6 UTC, 12 UTC and 18 UTC [MetCoOp Wiki, 2017]. The ECMWF-IFS forecasts are used 6 h prior to the actual cycle in MEPS. As synoptic observations are included in the model the snow-depth predictions underlay a special performance. Observations of snow-depth are only available at 6 UTC and 18 UTC, therefore is the snow analysis only performed twice daily [Homleid and Tveten, 2016, Müller et al., 2017].

4.2 MÈSONH

Meteo France [2009]; Seity et al. [2010]

4.3 SURFEX

Masson et al. [2013]

4.4 ICE3 SCHEME

Pinty and Jabouille [1998]

CHAPTER 5: DATA PROCESSING

The previous Chapters 3 and 4 represented the details within the forecast model outputs and the retrieved values. The following chapter will describe how the different variables were processed to achieve a comparison between the retrieved snow and the forecast model output.

5.1 LAYER THICKNESS IN MEPS

To compare the measurements from the surface with the MEPS data, the closest grid point is used to Haukeliseter.

MEPS has a vertical resolution in hybrid sigma pressure coordinates (0-1). The zero sigma level is at the top of the atmosphere, hence positive values are downward. To calculate the actual vertical pressure in Pa, a formula is provided in the OPeNDAP Dataset of `meps_full_2_5km_*.nc` by [Norwegian Meteorological Institute \[2016\]](#).

$$p(n, k, j, i) = ap(k) + b(k) \cdot ps(n, j, i) \quad [\text{Pa}]. \quad (5.1.1)$$

ps is the surface air pressure in Pa, and information about the variables ap , b are not given from the access form. [Maybe check in the documentation from BJ?](#)

The next step was to convert pressure-levels into actual heights by the use of the hypsometric equation. Here, the air temperature in model levels is used to calculate the mean temperature of each layer.

$$\bar{T} = \frac{\int_{p1}^{p2} T \partial \ln p}{\int_{p1}^{p2} \partial \ln p} \quad [\text{K}] \quad (5.1.2)$$

For the numerical integration, the Simpson rule was used, which is a build-in function in Python.

In the book from [Martin \[2006\]](#) steps of differentiating the hypsometric equation are presented by using the virtual temperature. But when the atmospheric mixing ratio is large, will the virtual temperature only be 1 % larger than the actual air temperature.

The thickness, Δz , of each layer is then be found by using the hypsometric equation from [Martin \[2006\]](#) and the previously calculated mean temperature (Equation (5.1.2)):

$$\Delta z = z_2 - z_1 = \frac{R_d \bar{T}}{g} \ln \left(\frac{p_1}{p_2} \right) \quad [\text{m}] \quad (5.1.3)$$

where R_d is gas constant for dry air with a value of $287 \text{ J kg}^{-1} \text{ K}^{-1}$, standard gravity $g = 9.81 \text{ m s}^{-2}$. p_1 and p_2 are the pressure levels at lower and higher levels, respectively ($p_2 < p_1$).

5.2 SNOW WATER CONTENT

To get a valid comparison between the SWC from the optimal estimation retrieval and the results from MEPS, the SWC is averaged over each hour. Taking the model initialisation of MEPS at 0 UTC the model produces forecast values at 0, 1, 2, ..., 22, 23, ..., 66 UTC. To approach hourly mean values from the retrieval SWC is averaged over 30 min prior and 29 min after each full hull hour. This leads to a match of the average value at the same time as from MEPS.

Since MEPS has a higher vertical resolution than the optimal estimation snowfall retrieval each vertical profile of SWC is averaged every 200 m. to accomplish the same vertical resolution only values above 100 m are used to start at the same range height as given from the MRR (Section 2.2).

Within the output from MEPS snow water content does not exist for each model layer. Hence the calculation of the SWC is performed by using the three solid precipitation categories given in MEPS. Namely the mixing ratio of snowfall (S), graupelfall (G) and the atmosphere cloud ice content (CIC). The mixing ratios are represented in kg kg^{-1} and a transformation to gm^{-3} is performed. Densities in each model level (ρ_{ml}) are calculated

and then multiplied with the sum of the solid precipitation mixing ratio.

$$\rho_{ml} = \frac{p_{ml}}{R_d T} \quad [\text{kg m}^{-3}] \quad (5.2.1)$$

$$SWC_{ml} = \rho_{ml} \cdot (S + G + CIC)_{ml} \cdot 10^6 \quad [\text{gm}^{-3}] \quad (5.2.2)$$

5.3 SNOW WATER PATH

The snow water path (SWP) is the vertically integrated value of the averaged SWC (Equations (3.4.1) and (5.2.2)), where the numerical Simpson's integration is applied.

$$\int_{h_0}^{h_1=3000\text{m}} \text{SWC}(h) dh \approx \frac{h_1 - h_0}{6} \left[\text{SWC}(h_0) + \text{SWC}(h_1) + 4\text{SWC}\left(\frac{h_0 + h_1}{2}\right) \right] \quad [\text{gm}^{-2}] \quad (5.3.1)$$

5.4 DEW POINT TEMPERATURE FOR SKEW-T LOG-P DIAGRAM

The Python module pyMeteo is used to calculate the dew point temperature of each ensemble member to study the stability of the atmosphere (<https://pythonhosted.org/pymeteo/>, last visited: 25.01.2018). The additional package thermo.py is able to calculate the dew point temperature if the pressure and the specific humidity in each level is known.

$$e_l = \ln \left(\frac{\frac{q_v}{\epsilon} \cdot \frac{p}{100}}{1 + \frac{q_v}{\epsilon}} \right) \quad (5.4.1)$$

$$T_d = 273.15 + \frac{243.5 \cdot e_l - 440.8}{19.48 - e_l} \quad [\text{K}] \quad (5.4.2)$$

where, q_v is the specific humidity, p pressure in [Pa], $\epsilon = R_d/R_v = 0.622$ with R_v gas constant for water vapour.

CHAPTER 6: ANALYSIS OF THE CHRISTMAS STORM 2016

The overleaf file from the Synoptic weather situation is here:

<https://www.overleaf.com/13946126xwbhxxrcwbyd>

This extreme weather event is chosen since MEPS just became operational and in addition surface snow instruments were installed to profile the vertical at Haukeliseter. Preliminary test showed the overestimation of snow accumulation by MEPS compared to the measurements at the ground. Furthermore, was changed the phase between the precipitation.

The next sections will give a definition of an extreme weather, a description of the different weather maps, and a synoptic interpretation of the storm and why mixed phase precipitation was observed at Haukeliseter.

Before the analysis each weather maps' purpose will be presented to understand the connections between them. For this, the ECMWF ...blabla product... so many grid degree ... is processed.

6.1 EXTREME WEATHER

'Extreme weather' is a meteorological term, associated with the extent of a weather type. The Norwegian Meteorological Institute declares an extreme event, if strong winds, large amounts of precipitation and large temperature changes are expected before the event occurs. As well as a large avalanche risk is present and coastal areas are influenced by extremely high-water levels. All this occurred during the Christmas storm [Olsen and

[Granerød, 2017\]](#).

Generally, an event is divided into four phases that it can be called extreme [[Pedersen and Rommetveit, 2013](#)].

Phase A: *Increased monitoring before the possible extreme weather.* The meteorologists give special attention to the weather situation. At this point it is not certain, that there will be an extreme weather event.

Phase B: *Short-term forecasts.* It is decided, that there will be an extreme event. The forecasts are more detailed and updates will be published at least every six hours. The event will get a name.

Phase C: *The extreme weather is in progress.* The meteorologists send out weather announcements at least every six hours.

Phase D: *The extreme weather event is over. Clean-up and repairs are in progress.* When the extreme weather is over the public is notified and information about the upcoming weather and clearing work is given.

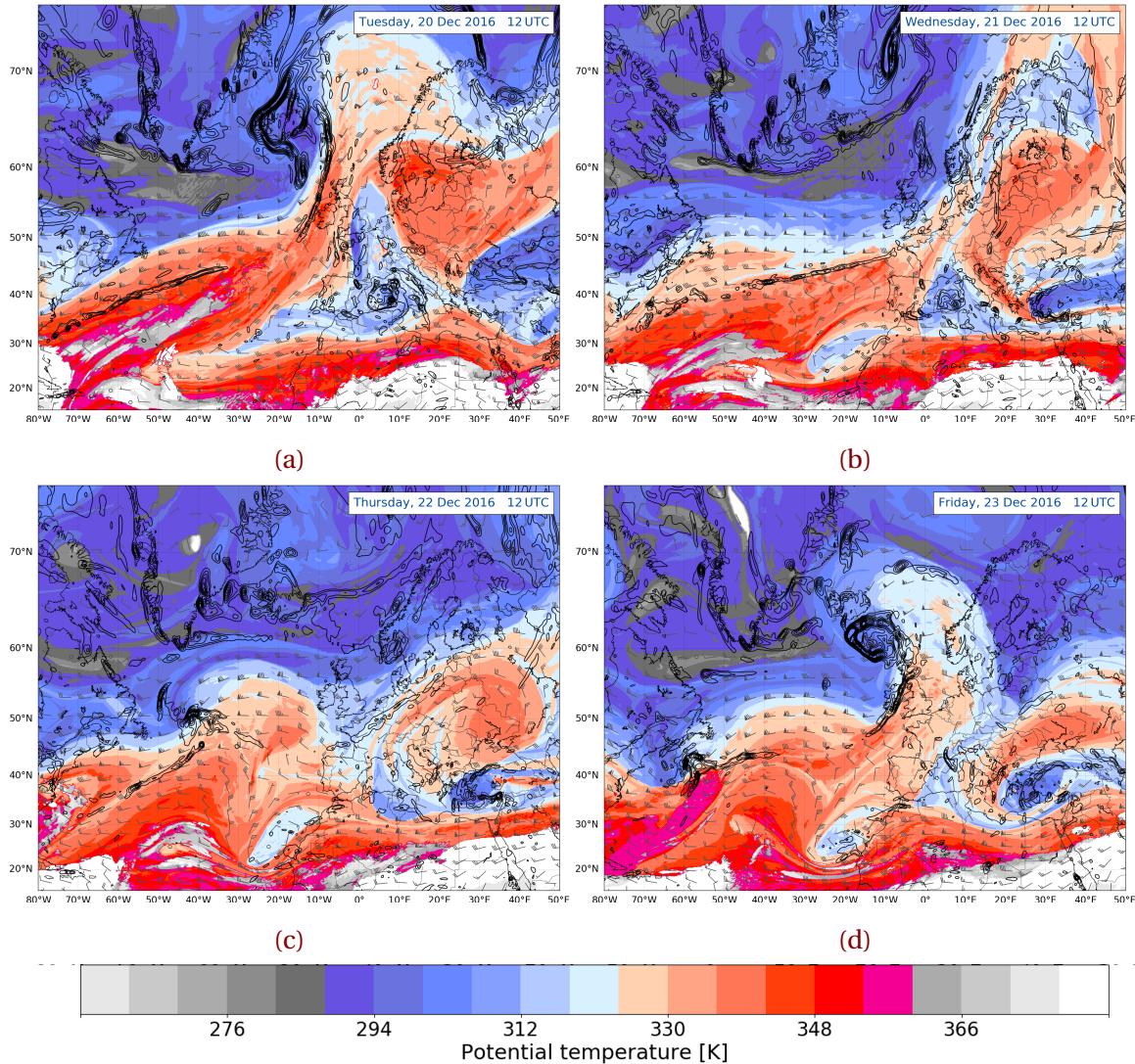
6.2 DYNAMIC TROPOPAUSE MAP

The dynamic tropopause map (DT), shown in Figure 6.2.1 presents the potential temperature distribution at the tropopause. Colder tropopause is associated with colder colours and vice versa warmer tropopause with warmer colours (shading according to the colorbar). Therefore, a warmer tropopause indicates an elevation of the atmospheric column.

The gradient at the 2 PVU (1 PV unit = $10^{-6} \text{ m}^2 \text{s}^{-1} \text{Kkg}^{-1}$; [Hoskins et al. \[1985\]](#)) surface, between the cold and warm area indicates the thermal wind. There is a slope between the cold and warm surfaces increasing towards the warmer column averaged temperature. An increased slope means also an increased pressure gradient force with increasing height and therefore an increase in geostrophic wind. This means, that there exists a vertical wind shear. From this, the mid-latitude jet stream can be pointed out. [do I need to present the equation of thermal wind?](#) Wind barbs in m s^{-1} indicate the direction of the wind flow, which is generally from west to east in the mid-latitudes.

The 925 – 850 hPa layer-averaged surface relative vorticity is shown in black contours, every $0.5 \times 10^{-4} \text{ s}^{-1}$. It represents the rotation of a fluid. [Does the relative vorticity need more explanation?](#)

Along the Rossby-Wave-Guide, troughs and ridges are seen which can be combined with the surface relative vorticity to understand the vertical dynamic interaction in the atmosphere. In case of a westward tilt between the surface cyclone and an upper level through an intensification of the surface cyclone is more likely to occur.



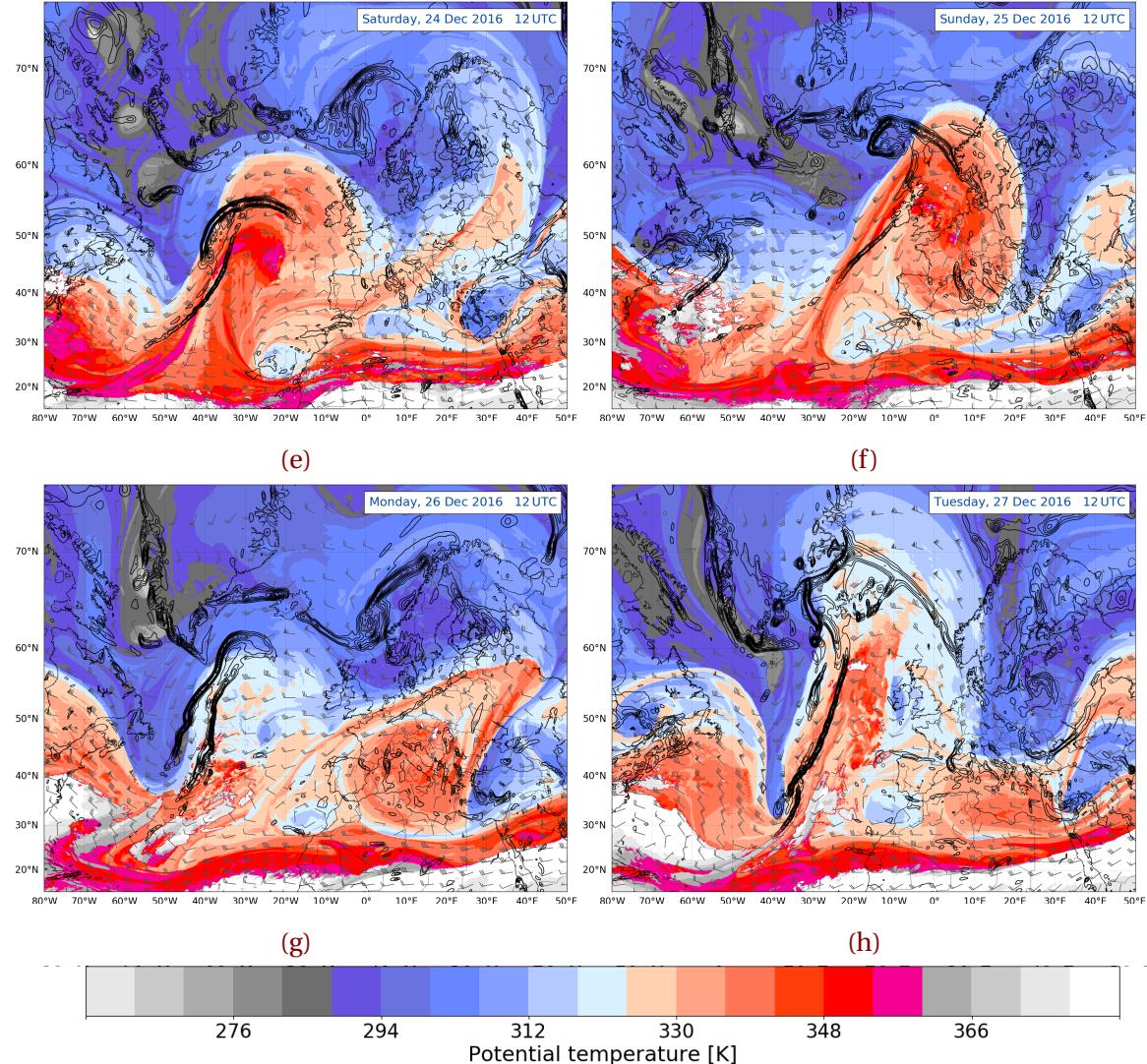
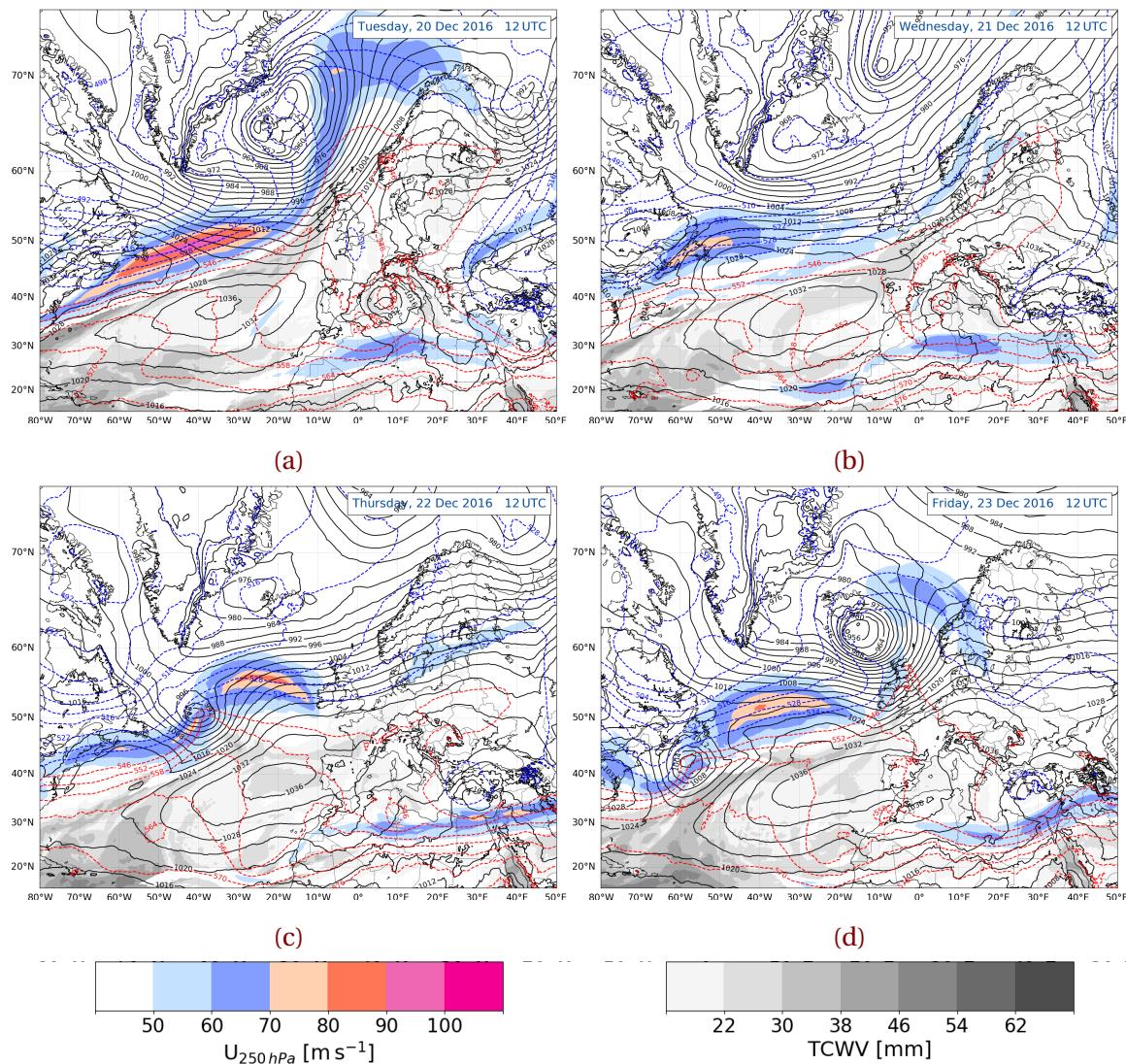


Figure 6.2.1: Dynamic tropopause analysis map, data from ECMWF at 2 PVU. During 20 December 2016 to 27 December 2016. Potential temperature [K] at the 2 PVU surface, shaded according to the colour bar. Total wind, barbs [m s^{-1}], and 925–850 hPa layer-averaged surface relative vorticity (black contours, every $0.5 \times 10^{-4} \text{ s}^{-1}$).

6.3 THICKNESS, SEA LEVEL PRESSURE, MOISTURE, AND WIND AT 250 hPa

A good overview gives the sea level pressure, 1000–500 hPa thickness map and winds at 250 hPa. Figure 6.3.1 shows, that it combines several important features of the vertical distribution within the atmosphere, for example.

Black contour lines indicate sea level pressure in hPa and makes it possible to observe cyclones and anticyclones at the sea surface.



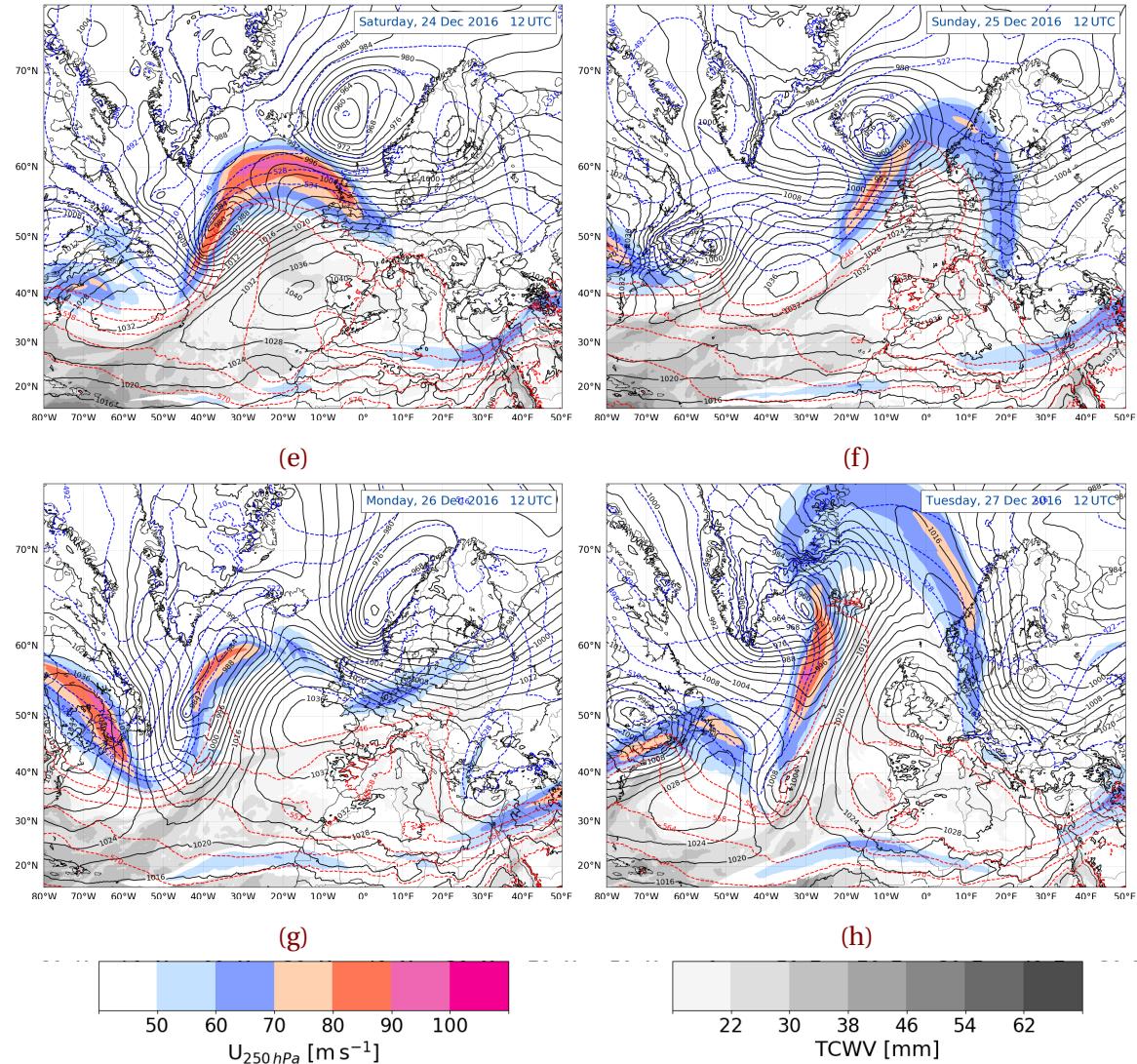


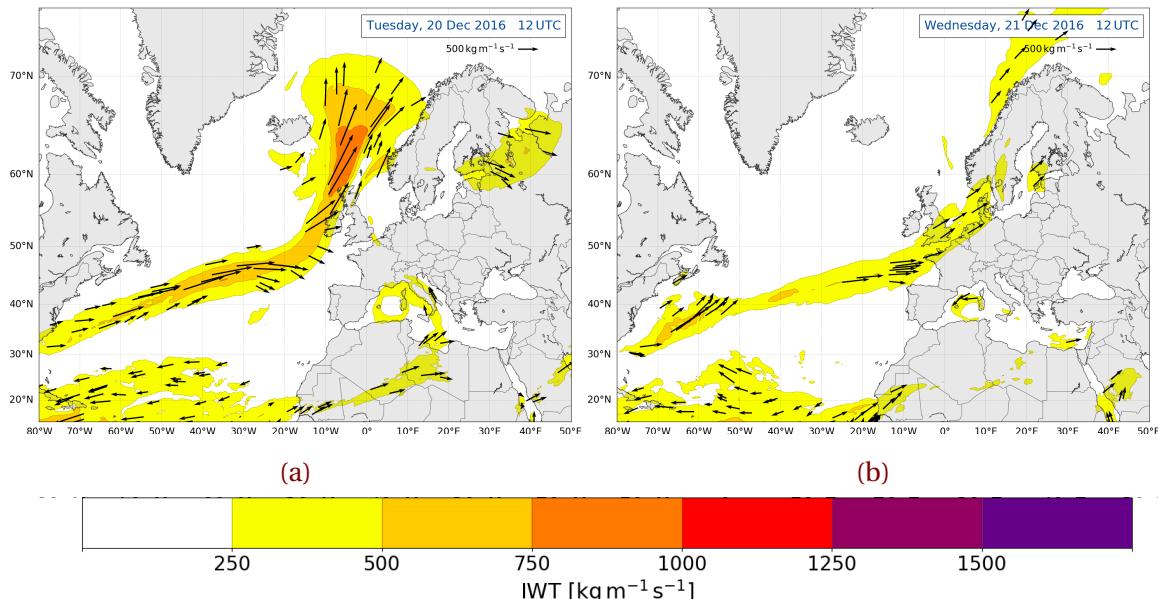
Figure 6.3.1: Jet, thickness, mean sea level pressure, and moisture synoptic analysis, data from ECMWF. During 20 December 2016 to 27 December 2016. 250 hPa wind speed, shaded according to the colour bar, [ms $^{-1}$]. 1000 – 500 hPa thickness, dashed contours every 6 dam, MSLP, black contours every 4 hPa, total column water vapour [mm], shaded according the grey scale.

The dashed, coloured contours show the vertical thickness between the 1000 hPa and 500 hPa surface, every 6 dam. The thickness between two pressure levels can be related to the hypsometric equation (Equation (5.1.3)). This is a relation of the mean temperature of the air between two pressure levels. Thus, high values of thickness mean relative warm, moist air (red, dashed). This can then be associated to rain or snow in mid-latitudes, depending on cold or warm air advection.

Gray shaded areas describe total precipitable water in the atmosphere in mm. It is an indicator for the amount of moisture to supply rainfall, and will be used to identify where moisture was present.

Colour shaded contours in Figure 6.3.1 indicate the mid-latitudal jet streaks at 250 hPa. Warmer colour is associated with higher wind speeds at this level.

6.4 ATMOSPHERIC RIVER MAP



An atmospheric river (AR) is a filament structure of intense moisture transport from the tropics to higher latitudes. Heavy precipitation can be associated with it, because the air is warm and moist. This can often be observed at mountain ranges at west coasts such as in Norway [include reference here](#). Due to orographic lifting will the moisture be released and follow high amounts of precipitation.

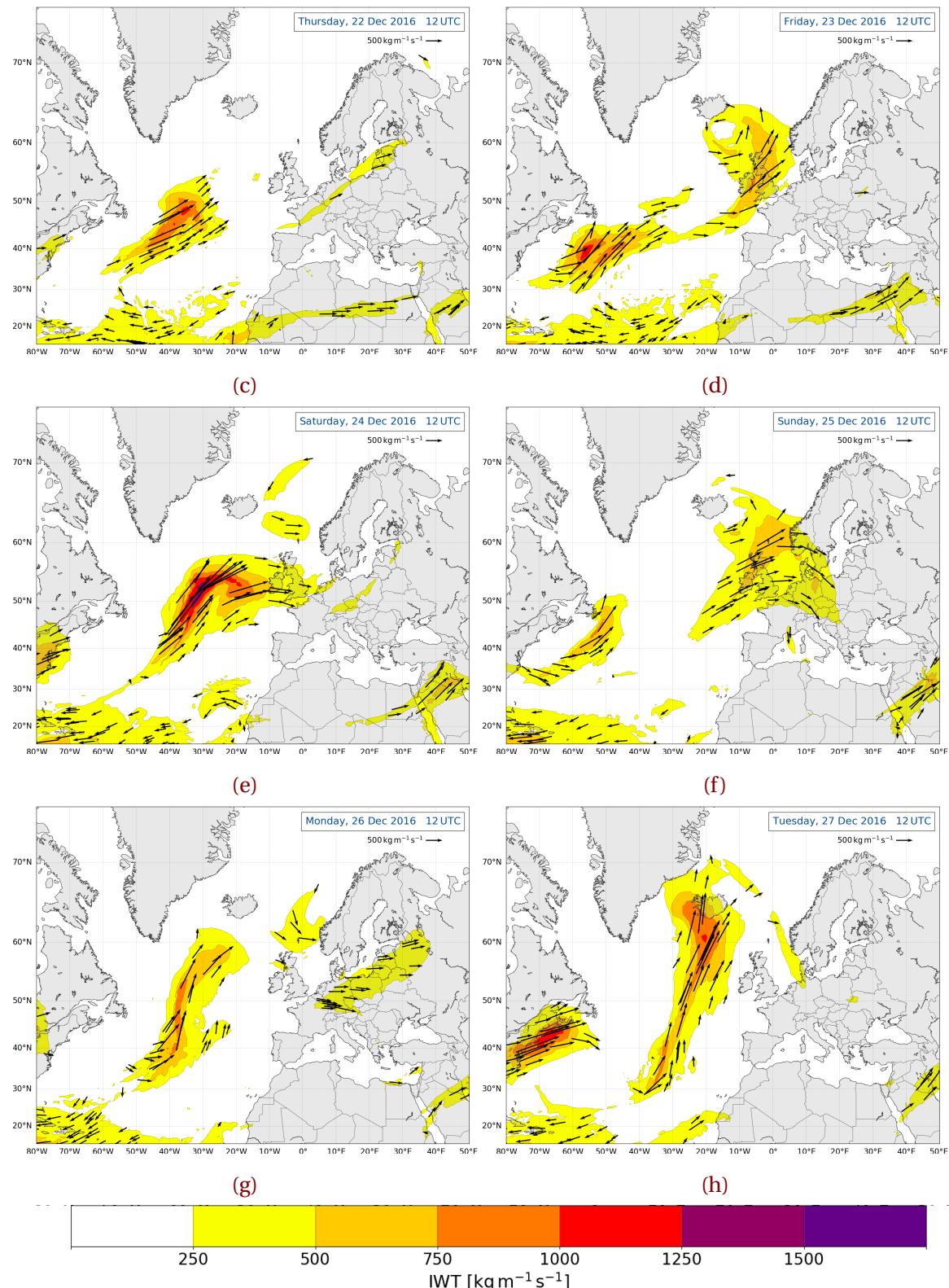


Figure 6.4.1: Atmospheric river analysis map, data from ECMWF. During 20 December 2016 to 27 December 2016. IVT, shaded according to the colour bar [$\text{kg m}^{-1} \text{s}^{-1}$]. Vectors, indicating the direction and magnitude of the IVT.

An atmospheric river is characterised if the integrated vapour transport shows values higher than $250 \text{ kg m}^{-1} \text{ s}^{-1}$ and is a continuous region larger than 2000 km [Rutz et al., 2014]. The integrated vapour transport (IVT) was calculated from the ECMWF data as followed:

$$IVT = \frac{1}{g} \int_{p_{sfc}}^{100 \text{ hPa}} q \mathbf{V} dp \quad [\text{kg m}^{-1} \text{ s}^{-1}] \quad (6.4.1)$$

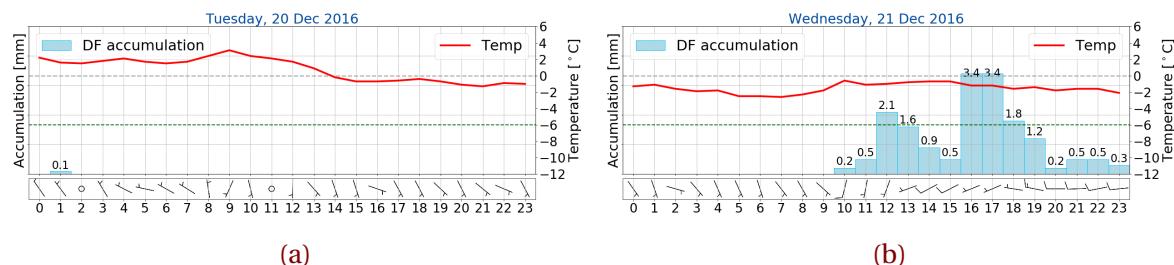
where g is the standard gravity, q the specific humidity, and \mathbf{V} the total wind vector at each pressure level p . The numerical, trapezoidal integration is performed by using data from the surface pressure p_{sfc} to 850 hPa in 50 hPa intervals and from 700 hPa to 100 hPa in 100 hPa intervals.

Figure 6.4.1 shows coloured contours of the integrated vapour transport (IVT) in $\text{kg m}^{-1} \text{ s}^{-1}$, where warmer colours indicate higher IVT. Stream vectors indicate the direction and intensity of the IVT flow.

6.5 OBSERVATIONS AT THE WEATHER MAST

The large scale synoptic analysis will be related to the local weather observations at Haukeliseter.

60 min accumulation is presented as bars in Figure 6.5.1 and will show the continuous precipitation at Haukeliseter during the extreme event. The possible change of precipitation will be investigated with the temperature. Snow fall is likely for temperatures up to 2°C . The intensity of the storm can be classified by the hourly averaged wind speed and direction as wind barbs in m s^{-1} . To understand which damage a storm can have, Færaas et al. [2016] released a table to associate wind strength with damage (see Table 6.5.1).



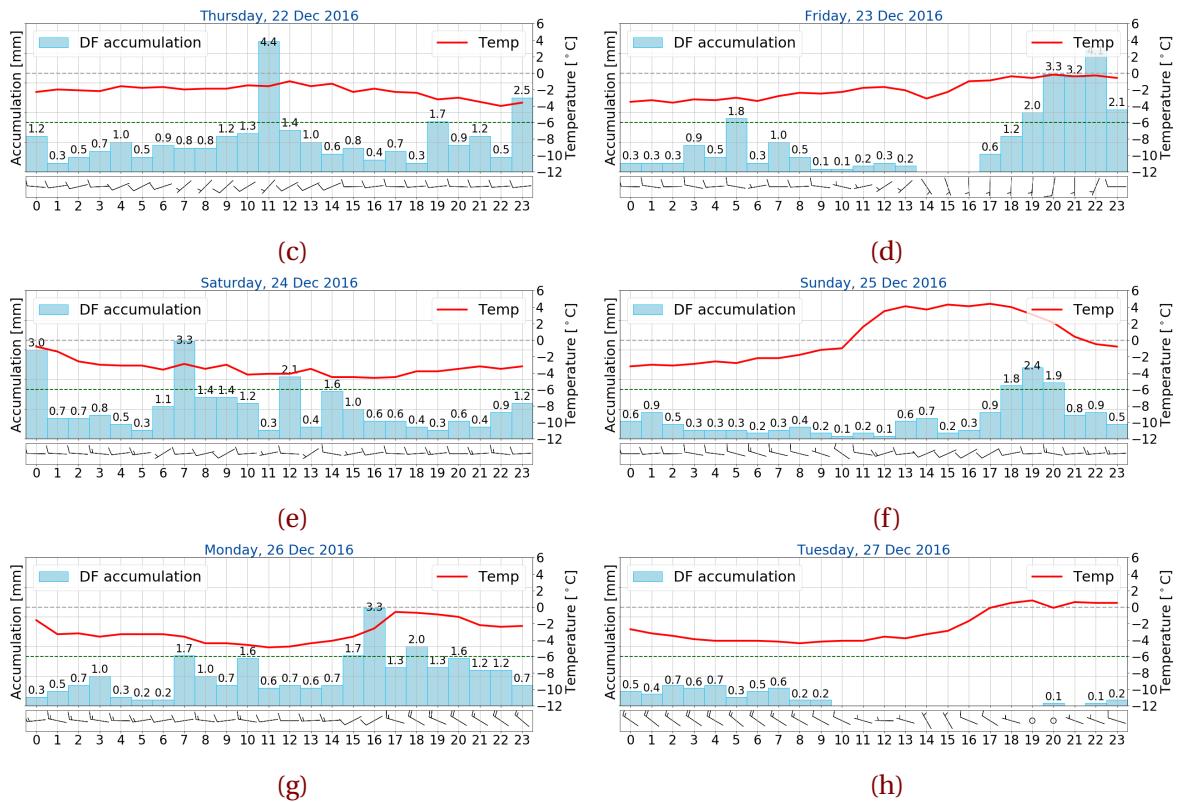


Figure 6.5.1: Observation from the weather mast at Haukeliseterd during 20 December 2016 to 27 December 2016. 60 min total accumulation [mm] in light blue as bar, temperature (red, $^{\circ}\text{C}$), and wind as barbs [m s^{-1}]. Gray dashed line indicates the freezing temperature and the green dashed line the 30-year climate mean temperature at -6°C . Hourly processed data taken from [eklima \[2016\]](#).

Table 6.5.1: Damage related to wind speed, from Færaas et al. [2016].

| | | |
|--------------|---|---|
| slight storm | $20.8 \text{ ms}^{-1} - 24.4 \text{ ms}^{-1}$ | Large trees sway and hiver. Roofs can blow down. |
| full storm | $24.5 \text{ ms}^{-1} - 28.4 \text{ ms}^{-1}$ | Trees are pulled up with clutter. Big damages to houses. |
| strong storm | $28.5 \text{ ms}^{-1} - 32.6 \text{ ms}^{-1}$ | Extensive damage. |
| hurricane | $>32.6 \text{ ms}^{-1}$ | Unusually large destruction. |

6.6 RADIOSONDE FROM STAVANGER

The Appendix A.2 includes images of vertical temperature profiles of Radiosondes from Stavanger, represented in a Skew-T log-P diagram. The vertical profiles extend from surface up to 300 hPa. (Note, that the representation of the vertical scale (in m) differs with each sounding from day to day.)

The vertical temperature distribution will identify the stability of the atmosphere and give information on convection and thickness of clouds. Furthermore, warm air and cold air advection can be identified by observing the change of wind with height.

Black and green line represent the atmospheric and dew-point temperature, respectively.

6.7 LARGE SCALE CIRCULATION

Everything has to get into relation with the Radiosondes, Figure A.2.1. How to do that? Any suggestions? They took me some time to make, so I want to include them!

21 December 2016

The dynamic tropopause map in Figure 6.2.1b shows that Norway is influenced by a change of elevated tropopause to a suppressed tropopause during 20 December 2016 to 21 December 2016. Hence the potential vorticity changed from positive to negative at the tropopause and cold air stretches right over Norway. A good amount of moisture is

transported from the low latitudes to high latitudes, influencing Norway's west coast. This can be seen in the surface maps (Figure 6.3.1b) as well as in the atmospheric river maps (Figure 6.4.1b). The westerly flow in Figure 6.3.1b is conducive to orographic lifting. The precipitation was probably snow when having a look at the moisture content and the cold air. The change from warm air to cold air can also be observed in the time series of temperature in Figure 6.5.1b. And the westerly flow, combined with a good amount of vapour transport from the tropics led to orographic lifting and precipitation at the Haukeliseter site. At around 60° W a formation of a cyclone at the baroclinic zone can be implied.

22 December 2016

Twenty-four hours later the analysis shows from 22 December 2016 phasing between the surface relative vorticity and the baroclinic zone at 50° N in the DT. The centre of the surface low is directly located below the temperature gradient at the 2 PVU surface, hence this is good for synoptic lifting. Furthermore, the strongest baroclinicity is observed on the south west side of the surface low. The synoptic map of the geopotential thickness and the surface pressure show the beginning of the frontal boundaries in Figure 6.3.1c. At the same time shows the AR map, Figure 6.4.1c, large values just at the baroclinic zone, where the low pressure is beginning to form. **Help?! Does that lead to even more lifting in this area? Or does it just mean that the cyclone gets a good amount of moisture?!** Norway is located in a cold area. The continues precipitation observed at Haukeliseter (Figure 6.5.1c) is associated with the westerly flow which is conducive to orographic lifting, and therefore moisture release.

23 December 2016

The begin of the ridging on the 22 December 2016 is more pronounced 24 h later. The warmer air pushes away the cold air, which covered Norway. The low pressure system moved north-east, and lies south of Iceland. The occluded front of this system passes through Haukeliseter, which is why a temperature 'jump' observed at 14 UTC. After this, Southern Norway is influenced by the warm sector, monitored as a temperature increase. The AR, as well as the total column water vapour amount in Figure 6.4.1d and Figure 6.3.1d,

respectively show the amount of moisture, transported from low latitudes.

At the same time forms a second cyclone at the baroclinic zone at 40° N. The atmospheric river map (Figure 6.4.1d) indicates a large amount of moisture at this latitude. Again, moist, warm air is conducive to intensify the surface cyclone. In addition, shows the DT map a phasing between the low level vorticity and the upper level baroclinic zone.

24 December 2016

After the passage of the cold front over Norway, Scandinavia is within colder air (compare Figure 6.2.1e and Figure 6.3.1e for 24 December 2016). Over the Atlantic warmer air starts to push the colder air northward. **something something with the low level vorticity and lifting; lifting at the right entrance region of the jet streak, and very high IVT.**

At Haukeliseter negative temperature up to -4° C is observed, compare Figure 6.5.1e. The westerly flow is again conducive for orographic lifting and associated precipitation.

25 December 2016

Twenty-four hours later the ridge is more pronounced and covers large parts of Norway. The surface low south-east of Iceland has build its frontal boundaries, which can be seen in the low-level vorticity of Figure 6.2.1f. The warm front lies west of Haukeliseter and starts to be observed at the measurement site (compare Figure 6.5.1f for 25 December 2016). Figure 6.4.1f indicating the integrated vapour transport shows that a lot of moisture is transported from the Atlantic, towards Great Britain and south-western Norway. Together with the lifting at the surface boundary a sufficient amount of precipitation is observed. Since the ridging brings more moist (Figure 6.4.1f), warm air (Figure 6.2.1f) and Norway lies in a warm sector (Figure 6.3.1f) the assumption will be made, that the precipitation changed from solid to liquid.

26 December 2016

Within the next twenty-four hours the cold front passed through (temperature change in Figure 6.5.1g for 25 December 2016 to 26 December 2016). Norway is covered in cold air (Figure 6.2.1g). The surface low-level indicates the occlusion of the cyclone and therefore a weakening. The wind is still from the west which is helpful for orographic lifting. The

moisture content is still present but much weaker and smaller in extend. Since Norway is covered in cold air, the temperature is below zero and the precipitation had to be solid.

27 December 2016

The images of 27 December 2016 show that the storm passed and disappeared. Southern Norway lies in cold air (Figure 6.2.1h), but on the right exit region of the jet (\rightarrow sinking motion of cold air), compare 6.3.1h. A small amount of moisture is present (Figure 6.4.1h). Because of the wind change from west to north-west follows that orographic lifting is not present and the precipitation amount decreases at the end of the storm.

CHAPTER 7: RESULTS

The overleaf file from the Results is here:

<https://www.overleaf.com/15070250fhqztvnygzmn>

7.1 WEDNESDAY, 21 December 2016

7.2 THURSDAY, 22 December 2016

7.3 FRIDAY, 23 December 2016

7.4 SATURDAY, 24 December 2016

7.5 SUNDAY, 25 December 2016

7.6 MONDAY, 26 December 2016

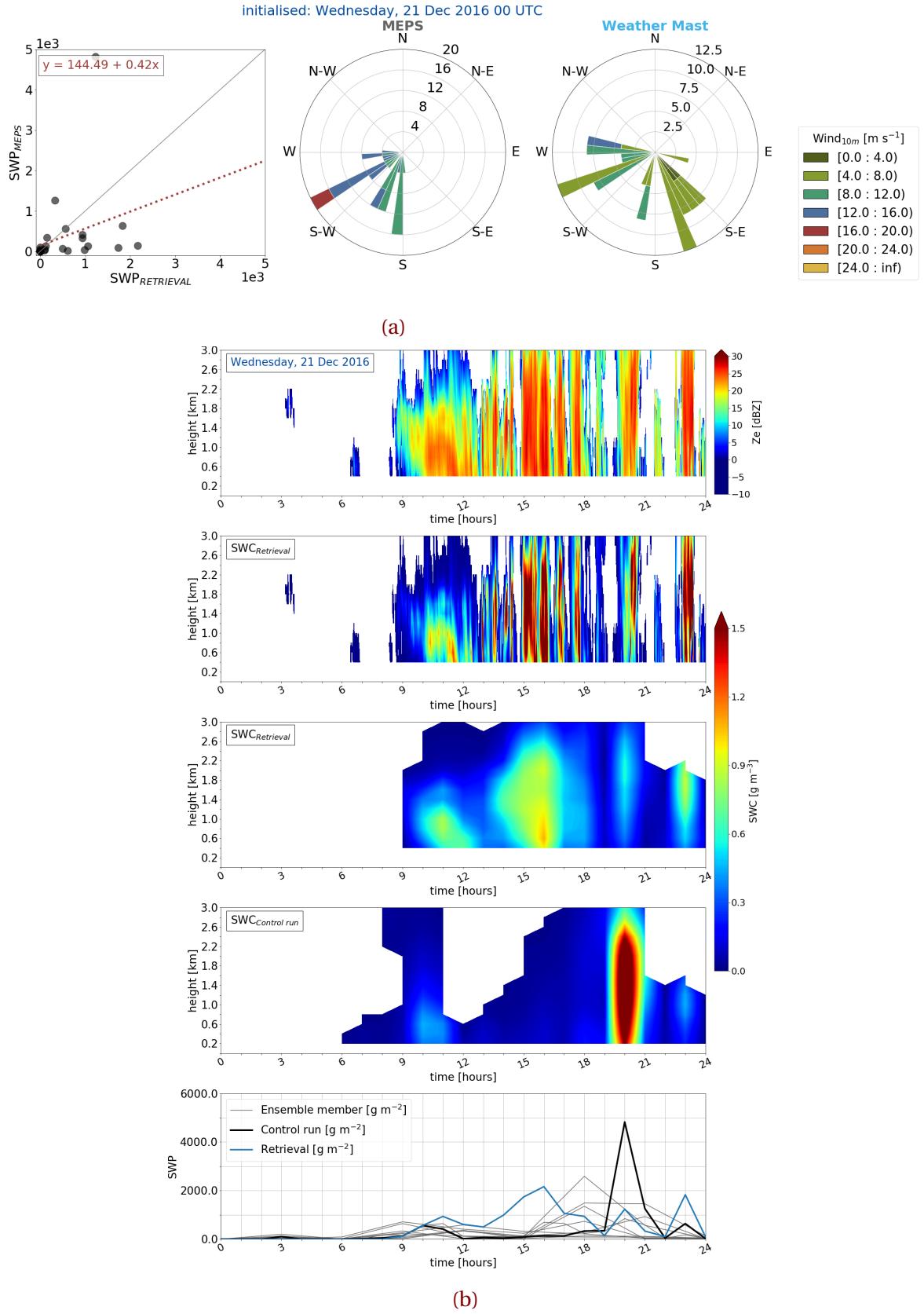


Figure 7.1.1

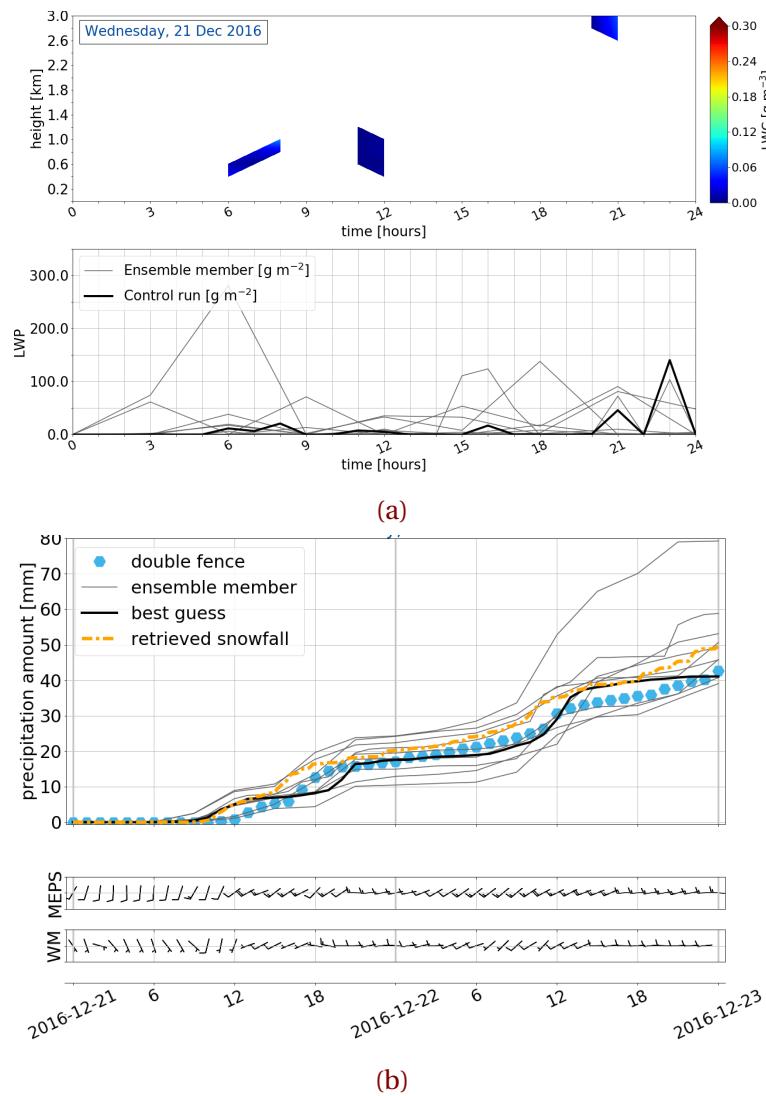


Figure 7.1.2

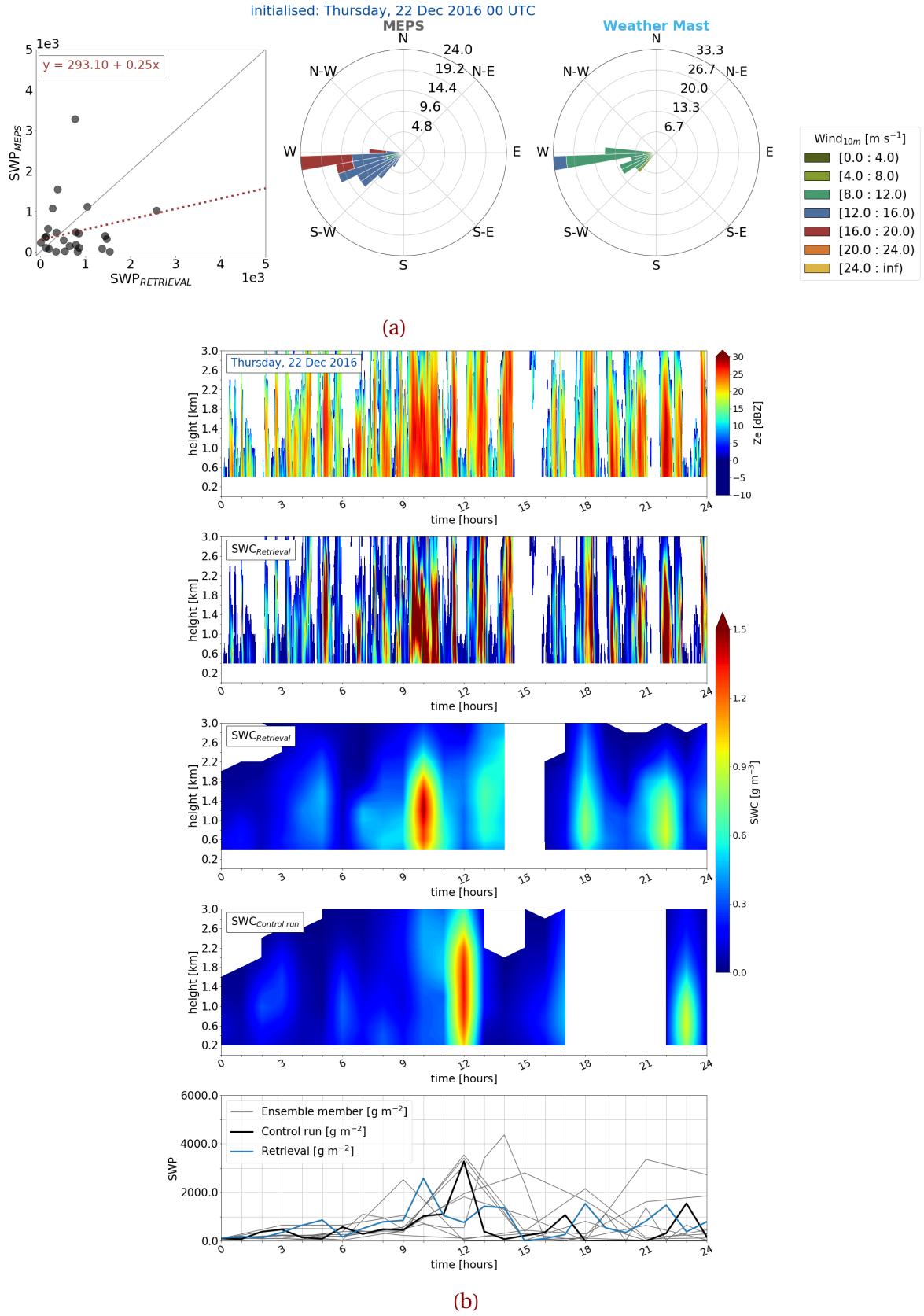


Figure 7.2.1

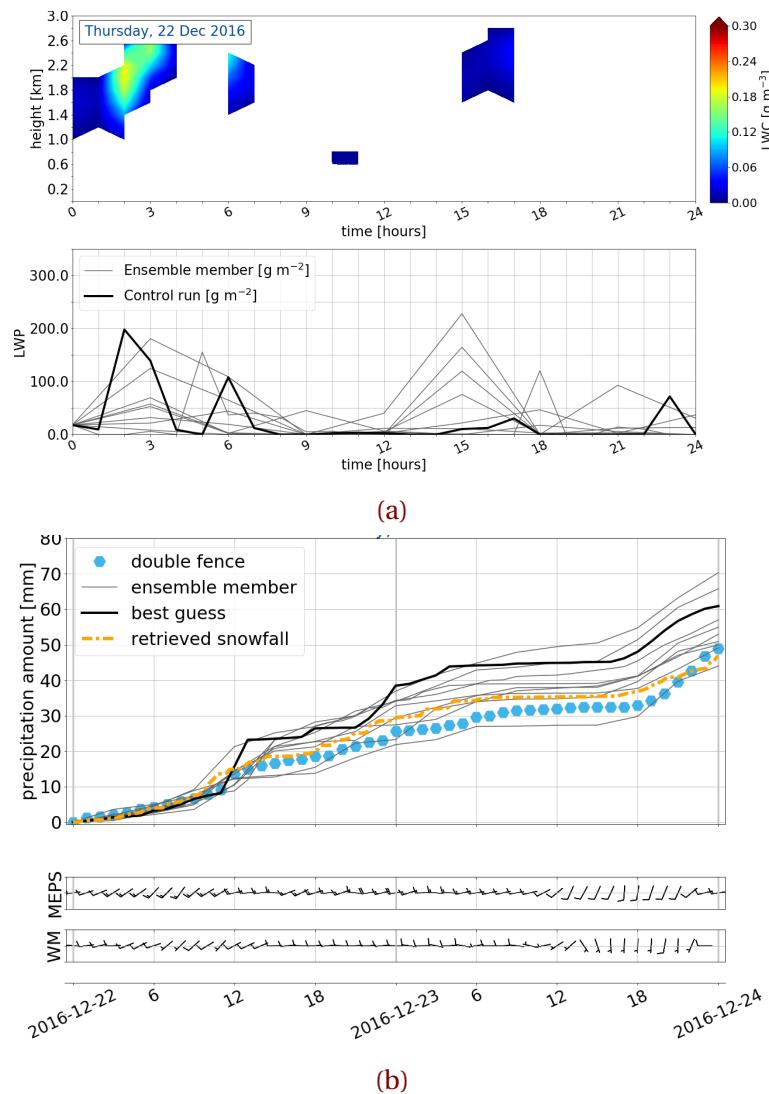


Figure 7.2.2

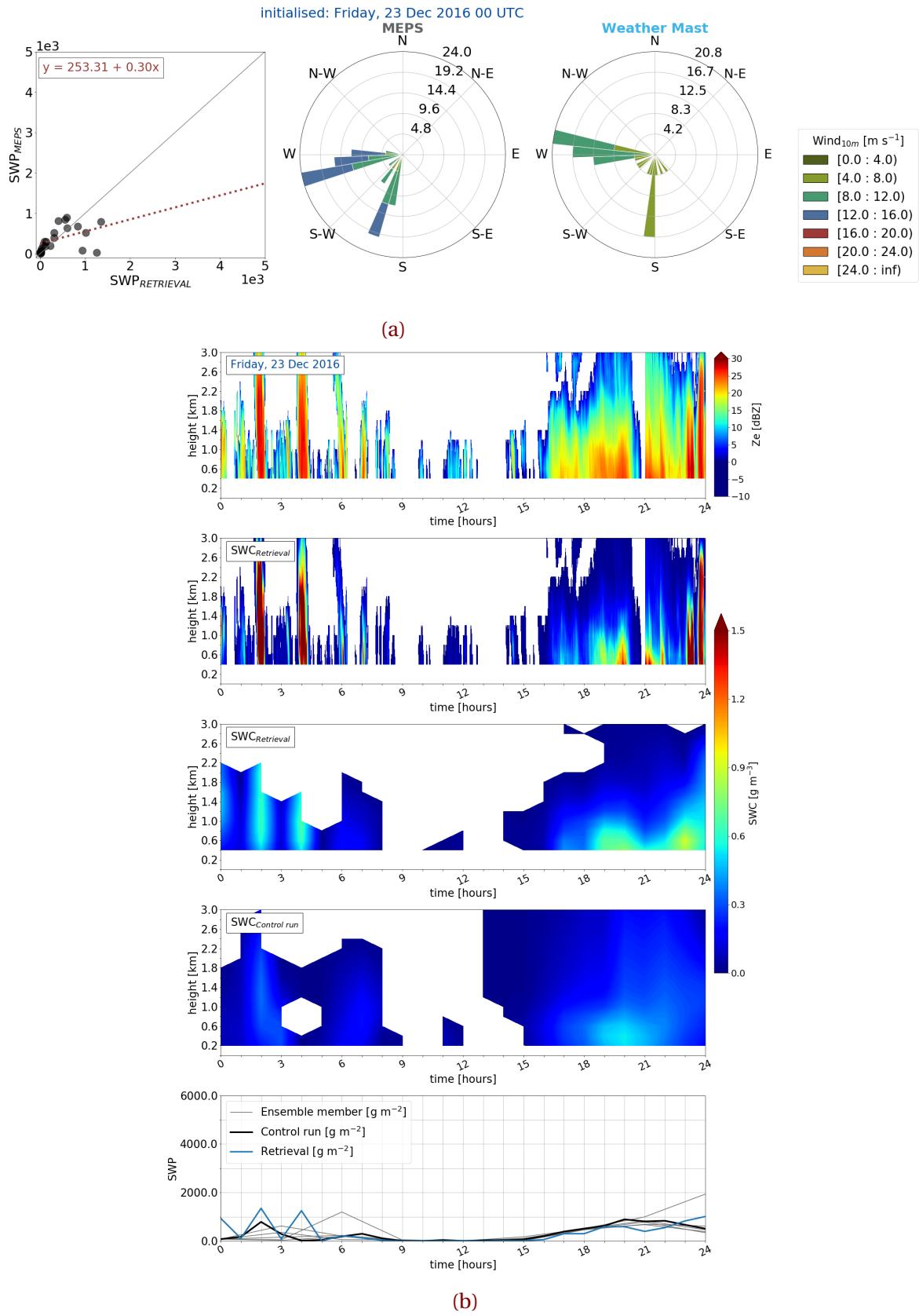


Figure 7.3.1

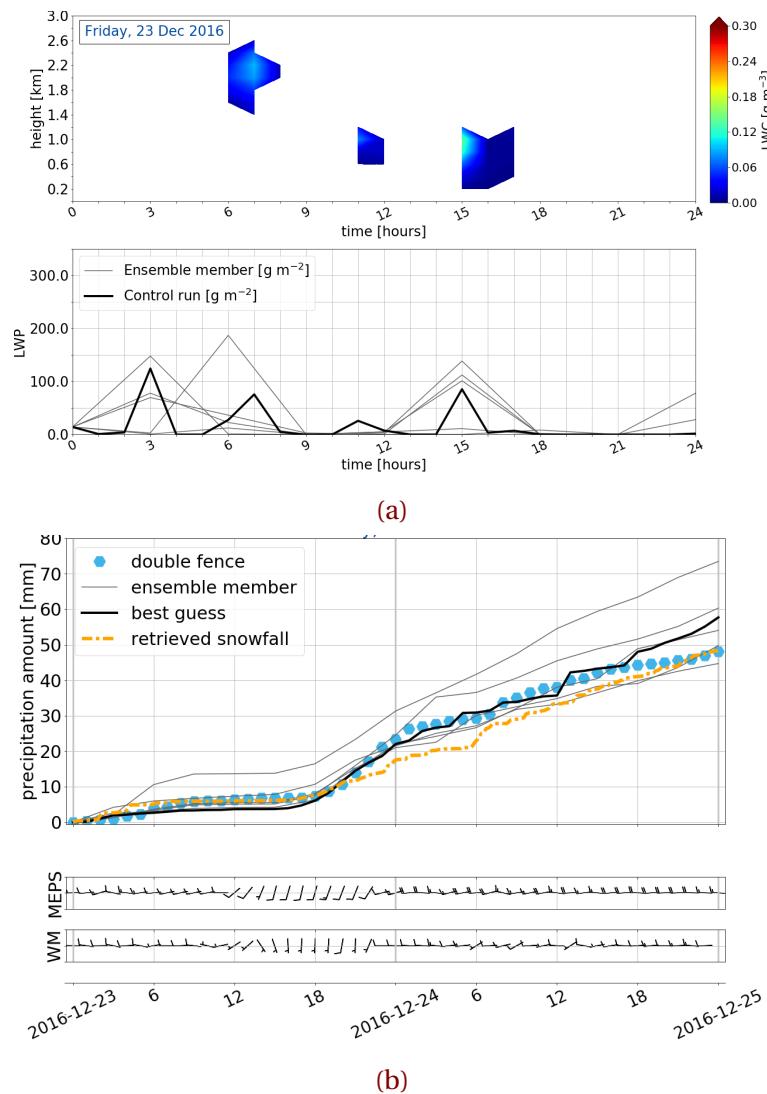


Figure 7.3.2

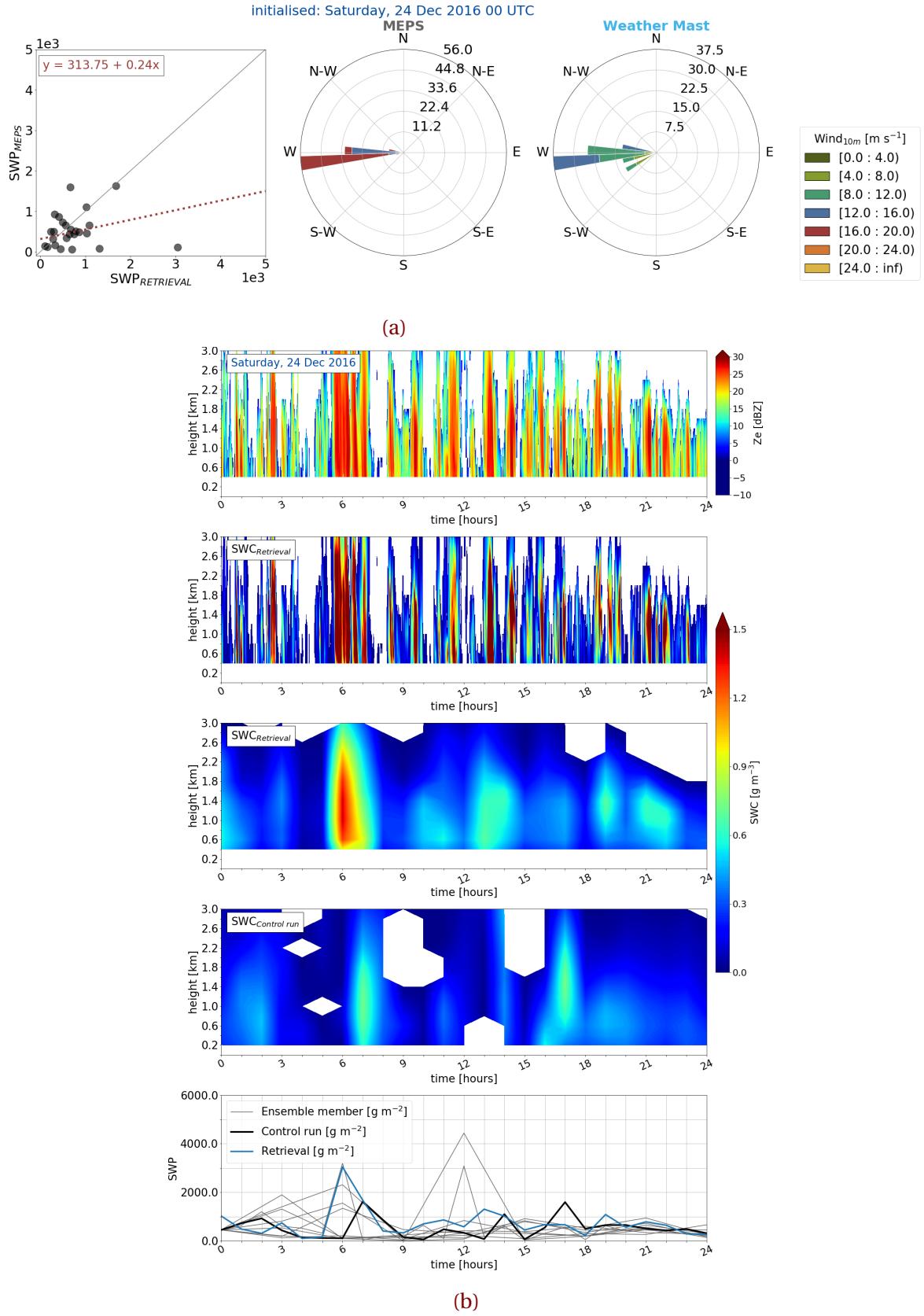


Figure 7.4.1

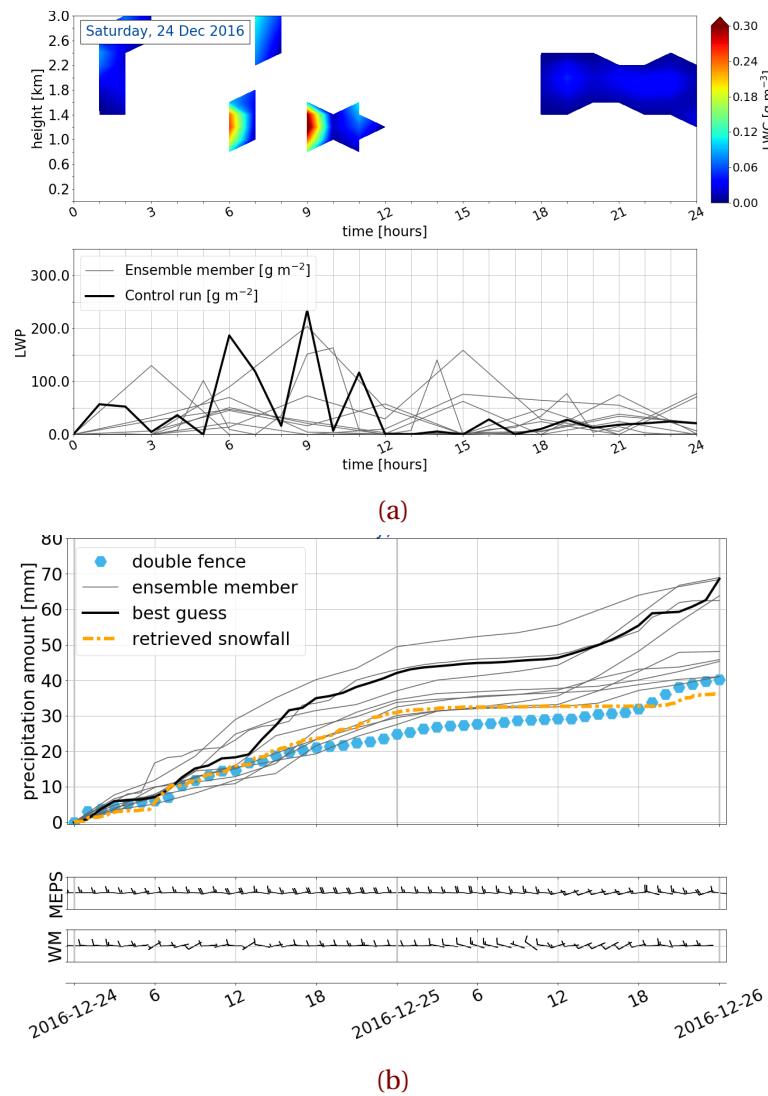


Figure 7.4.2

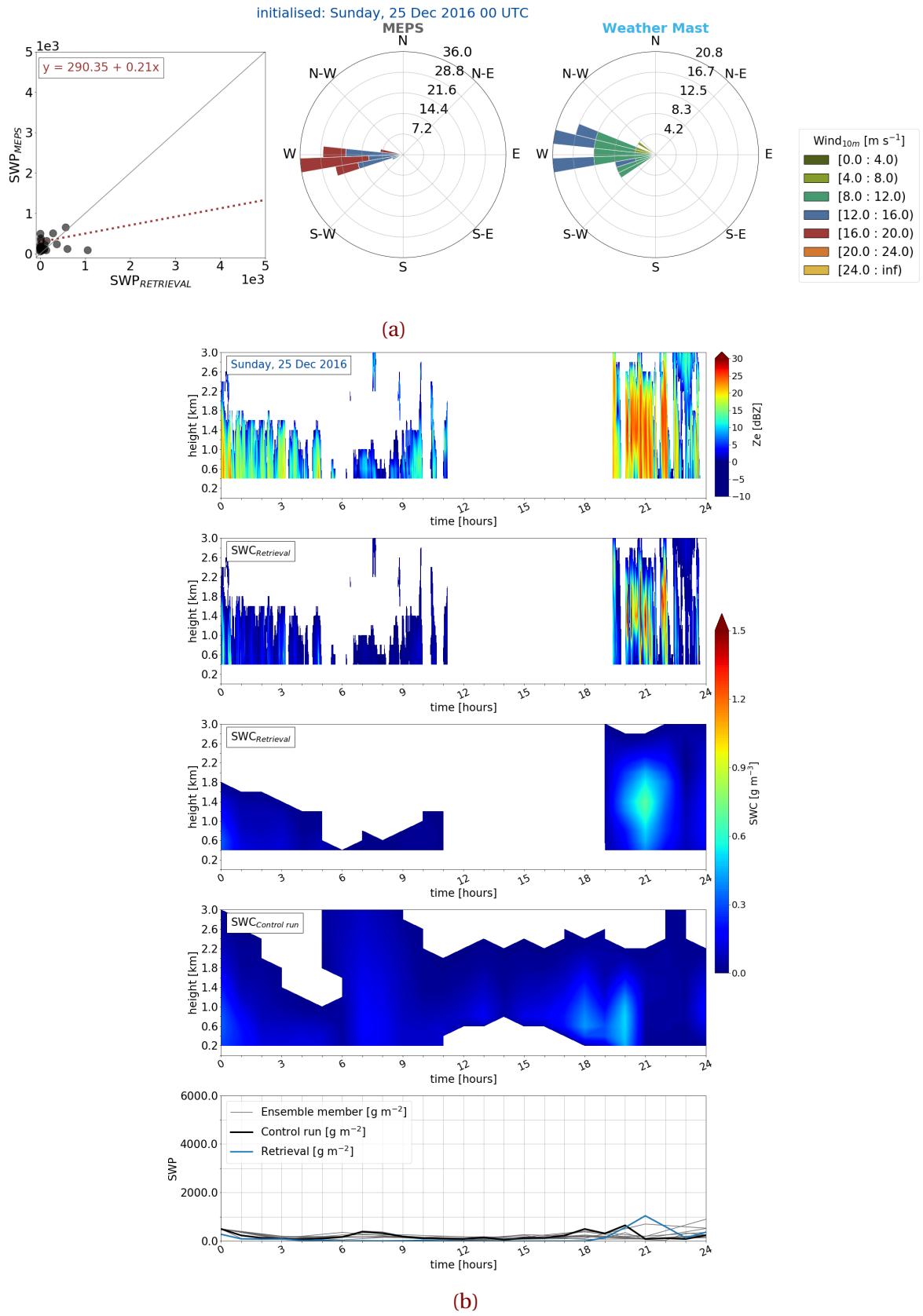


Figure 7.5.1

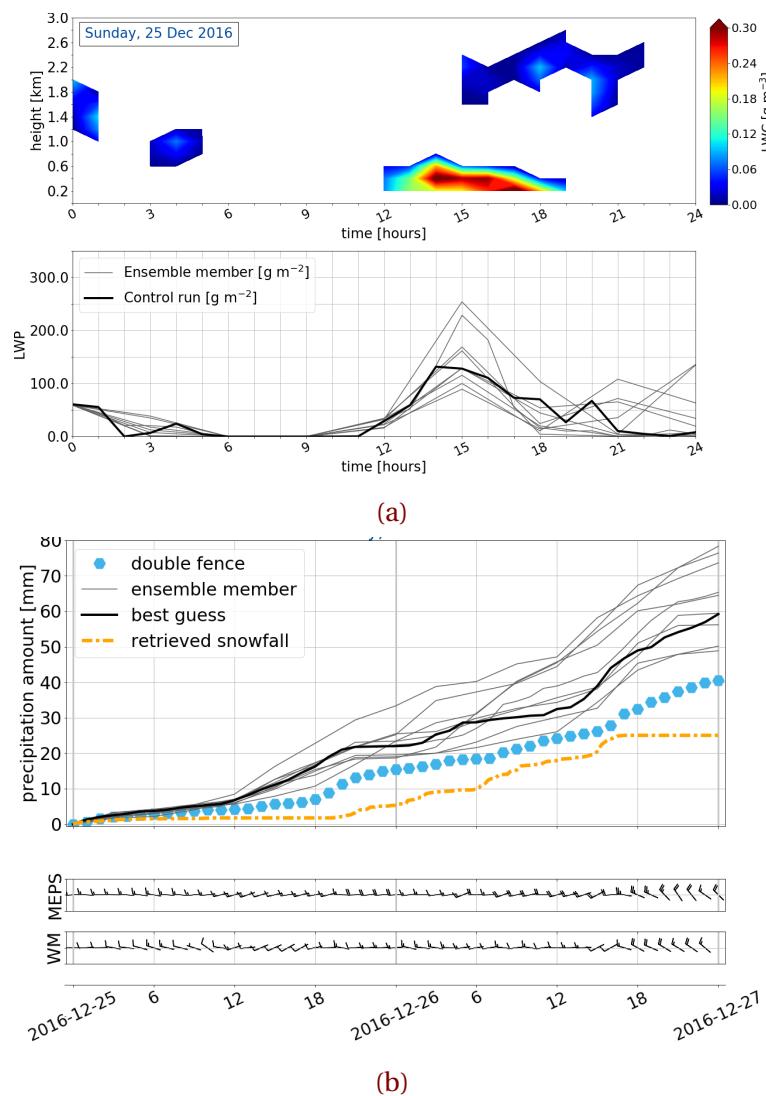


Figure 7.5.2

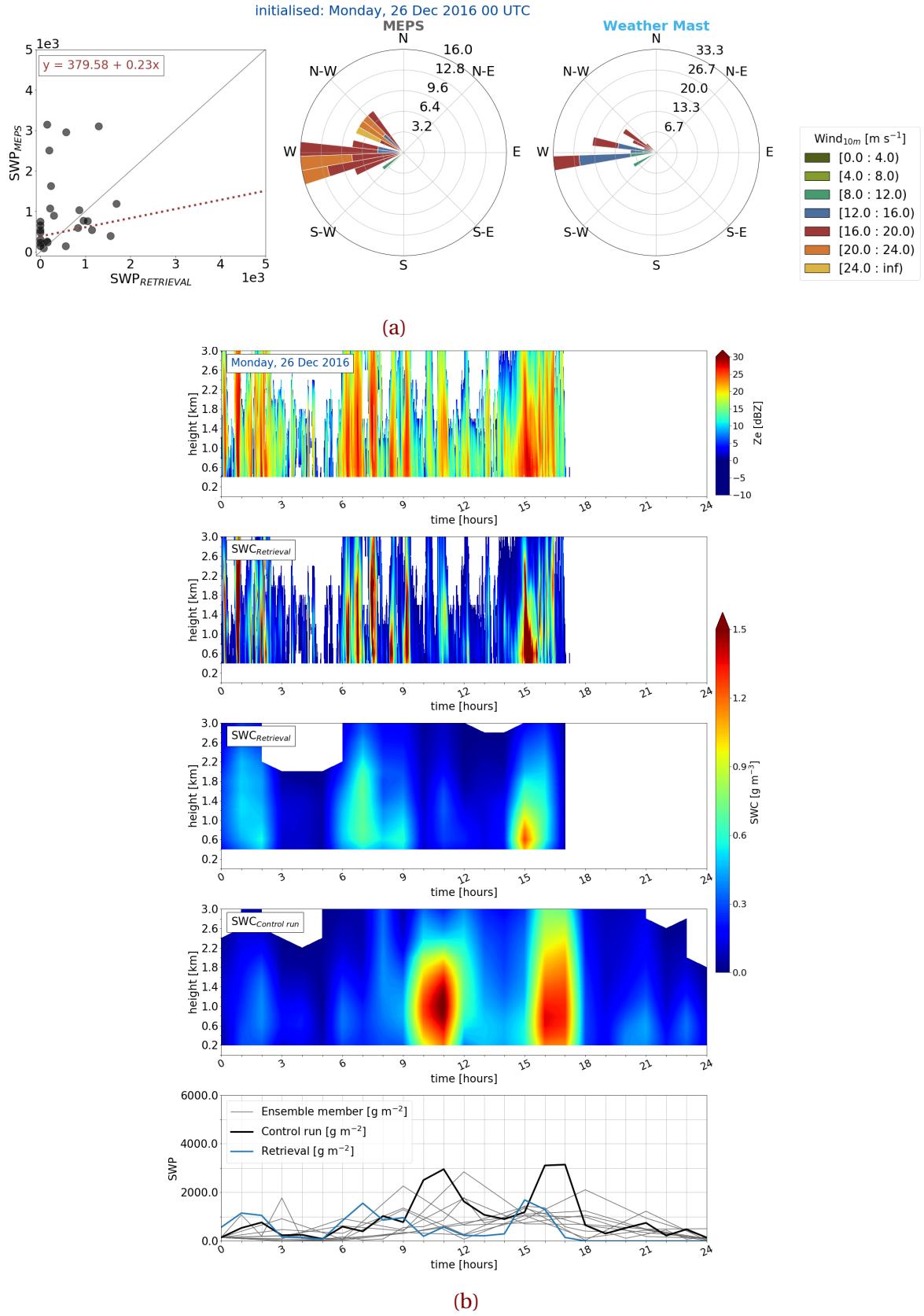


Figure 7.6.1

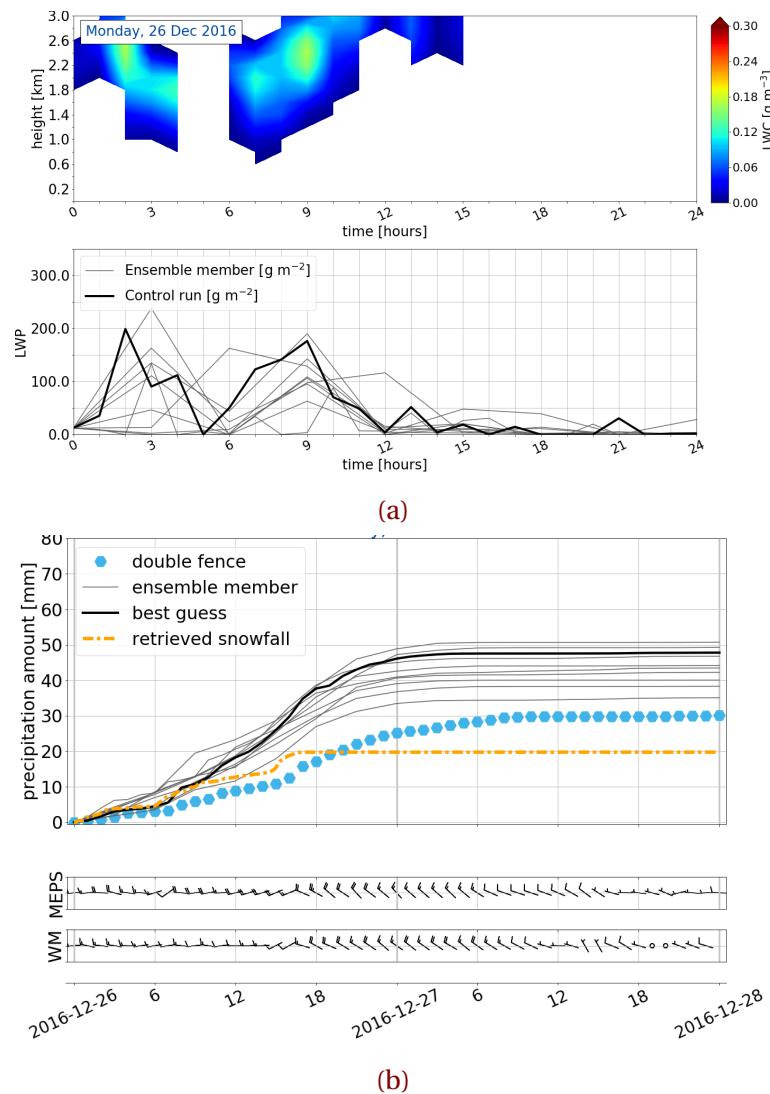


Figure 7.6.2

Table 7.6.1: Maximum values of the snow water and liquid water content from the retrieval and MEPS

| | SWC [kg m ⁻³] | HEIGHT [m] | TIME | LWC [kg m ⁻³] | HEIGHT [m] | TIME |
|-------------------------|------------------------------|---------------|--------|------------------------------|---------------|--------|
| Wed, 21 Dec 2016 | | | | | | |
| RETRIEVAL | 1.08 | 600.0 | 16 UTC | | | |
| MEPS | 2.11 | 1400.0 | 20 UTC | 0.15 | 2200. 0 | 23 UTC |
| Thu, 22 Dec 2016 | | | | | | |
| RETRIEVAL | 1.46 | 1200.0 | 10 UTC | | | |
| MEPS | 1.35 | 1400.0 | 12 UTC | 0.20 | 2000.0 | 02 UTC |
| Fri, 23 Dec 2016 | | | | | | |
| RETRIEVAL | 0.91 | 600.0 | 23 UTC | | | |
| MEPS | 0.54 | 400.0 | 20 UTC | 0.14 | 1000.0 | 15 UTC |
| Sat, 24 Dec 2016 | | | | | | |
| RETRIEVAL | 1.39 | 1000.0 | 06 UTC | | | |
| MEPS | 0.73 | 1400.0 | 17 UTC | 0.33 | 1200.0 | 09 UTC |
| Sun, 25 Dec 2016 | | | | | | |
| RETRIEVAL | 0.69 | 1400.0 | 21 UTC | | | |
| MEPS | 0.50 | 800.0 | 20 UTC | 0.34 | 200.0 | 17 UTC |
| Mon, 26 Dec 2016 | | | | | | |
| RETRIEVAL | 1.25 | 600.0 | 15 UTC | | | |
| MEPS | 1.55 | 1000.0 | 11 UTC | 0.17 | 2400.0 | 09 UTC |

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APPENDIX A:

A.1

Table A.1.1: Branched 6-arm spatial particle with porosities, 2D, mass oriented scattering scheme at 24.0GHz. D, particle size of the snow particle; m(D), particle mass; $\sigma_{\text{bk}}(\mathbf{D})$ and $\sigma_{\text{ext}}(\mathbf{D})$, backscattering and extinction cross-section, respectively.

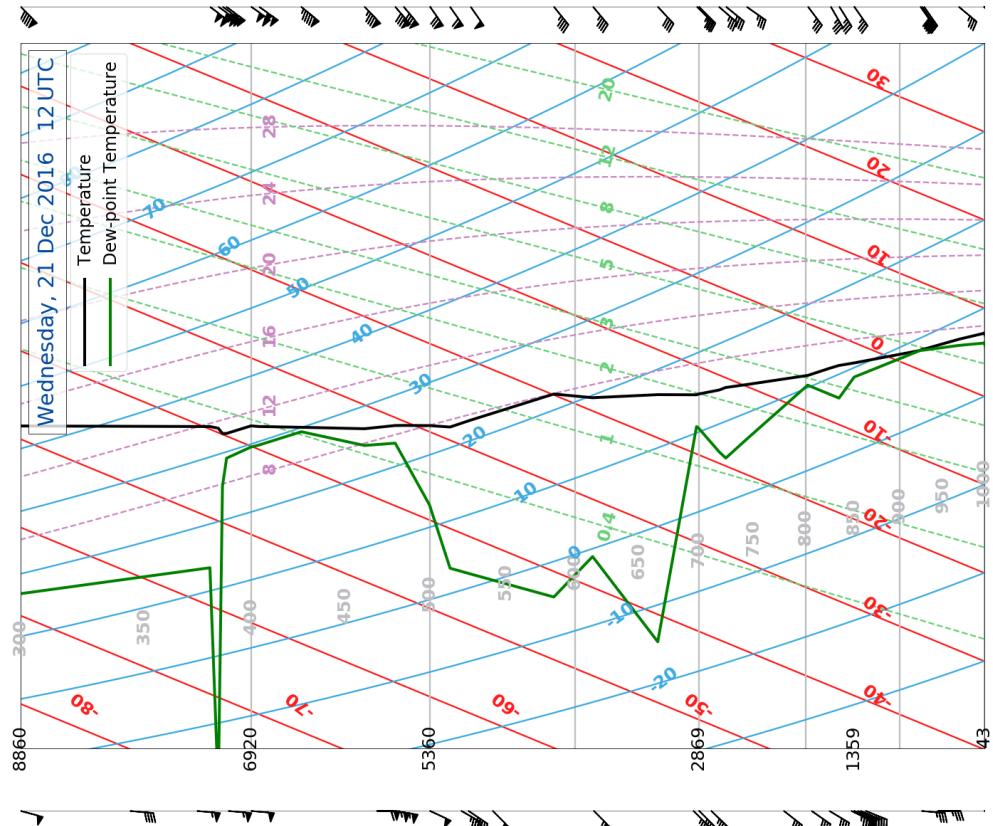
| D [μm] | m(D) [kg] | $\sigma_{\text{bk}}(\mathbf{D})$ [m ⁻²] | $\sigma_{\text{ext}}(\mathbf{D})$ [m ⁻²] | D [μm] | m(D) [kg] | $\sigma_{\text{bk}}(\mathbf{D})$ [m ⁻²] | $\sigma_{\text{ext}}(\mathbf{D})$ [m ⁻²] |
|-----------|-----------------------------|--|---|-----------|----------------------------|--|---|
| 35.27 | 1.68529 × 10 ⁻¹⁰ | 8.85111 × 10 ⁻¹⁷ | 4.85381 × 10 ⁻¹⁷ | 428.20 | 3.01577 × 10 ⁻⁷ | 1.7526 × 10 ⁻¹⁰ | 5.67393 × 10 ⁻¹⁰ |
| 41.73 | 2.79128 × 10 ⁻¹⁰ | 2.00612 × 10 ⁻¹⁶ | 1.28776 × 10 ⁻¹⁶ | 463.00 | 3.81242 × 10 ⁻⁷ | 3.58177 × 10 ⁻¹⁰ | 8.76374 × 10 ⁻¹⁰ |
| 47.87 | 4.21355 × 10 ⁻¹⁰ | 4.792 × 10 ⁻¹⁶ | 2.9959 × 10 ⁻¹⁶ | 496.90 | 4.71265 × 10 ⁻⁷ | 5.86279 × 10 ⁻¹⁰ | 1.30417 × 10 ⁻⁹ |
| 53.76 | 5.96809 × 10 ⁻¹⁰ | 1.02733 × 10 ⁻¹⁵ | 6.08871 × 10 ⁻¹⁶ | 530.10 | 5.72178 × 10 ⁻⁷ | 8.50141 × 10 ⁻¹⁰ | 1.8862 × 10 ⁻⁹ |
| 59.45 | 8.07074 × 10 ⁻¹⁰ | 1.68272 × 10 ⁻¹⁵ | 1.09633 × 10 ⁻¹⁵ | 562.60 | 6.84002 × 10 ⁻⁷ | 1.04566 × 10 ⁻⁹ | 2.60854 × 10 ⁻⁹ |
| 70.34 | 1.3368 × 10 ⁻⁹ | 5.7444 × 10 ⁻¹⁵ | 3.61096 × 10 ⁻¹⁵ | 594.50 | 8.07074 × 10 ⁻⁷ | 1.54514 × 10 ⁻⁹ | 3.68176 × 10 ⁻⁹ |
| 80.69 | 2.01798 × 10 ⁻⁹ | 1.0899 × 10 ⁻¹⁴ | 6.93961 × 10 ⁻¹⁵ | 625.80 | 9.41379 × 10 ⁻⁷ | 1.61704 × 10 ⁻⁹ | 4.48578 × 10 ⁻⁹ |
| 90.63 | 2.85939 × 10 ⁻⁹ | 2.244 × 10 ⁻¹⁴ | 1.42249 × 10 ⁻¹⁴ | 656.60 | 1.08733 × 10 ⁻⁶ | 2.10709 × 10 ⁻⁹ | 5.71184 × 10 ⁻⁹ |
| 100.20 | 3.86421 × 10 ⁻⁹ | 3.7814 × 10 ⁻¹⁴ | 2.73019 × 10 ⁻¹⁴ | 687.00 | 1.24546 × 10 ⁻⁶ | 3.31567 × 10 ⁻⁹ | 7.85938 × 10 ⁻⁹ |
| 109.50 | 5.04313 × 10 ⁻⁹ | 7.05869 × 10 ⁻¹⁴ | 5.36211 × 10 ⁻¹⁴ | 717.00 | 1.41584 × 10 ⁻⁶ | 3.73598 × 10 ⁻⁹ | 9.50817 × 10 ⁻⁹ |
| 118.60 | 6.40785 × 10 ⁻⁹ | 1.16874 × 10 ⁻¹³ | 9.74644 × 10 ⁻¹⁴ | 746.50 | 1.59789 × 10 ⁻⁶ | 4.40591 × 10 ⁻⁹ | 1.14824 × 10 ⁻⁸ |
| 127.40 | 7.94266 × 10 ⁻⁹ | 1.67227 × 10 ⁻¹³ | 1.56602 × 10 ⁻¹³ | 775.60 | 1.79214 × 10 ⁻⁶ | 5.1432 × 10 ⁻⁹ | 1.37371 × 10 ⁻⁸ |
| 144.50 | 1.15894 × 10 ⁻⁸ | 3.41952 × 10 ⁻¹³ | 4.19048 × 10 ⁻¹³ | 804.40 | 1.99928 × 10 ⁻⁶ | 4.21261 × 10 ⁻⁹ | 1.59603 × 10 ⁻⁸ |
| 160.90 | 1.60002 × 10 ⁻⁸ | 7.30397 × 10 ⁻¹³ | 1.05187 × 10 ⁻¹² | 832.90 | 2.2194 × 10 ⁻⁶ | 7.0875 × 10 ⁻⁹ | 1.90438 × 10 ⁻⁸ |
| 176.80 | 2.12278 × 10 ⁻⁸ | 1.13638 × 10 ⁻¹² | 2.00359 × 10 ⁻¹² | 861.10 | 2.45255 × 10 ⁻⁶ | 7.606 × 10 ⁻⁹ | 2.17023 × 10 ⁻⁸ |
| 192.30 | 2.73147 × 10 ⁻⁸ | 2.05333 × 10 ⁻¹² | 3.63531 × 10 ⁻¹² | 888.90 | 2.69784 × 10 ⁻⁶ | 9.61605 × 10 ⁻⁹ | 2.52476 × 10 ⁻⁸ |
| 236.50 | 5.08103 × 10 ⁻⁸ | 5.94138 × 10 ⁻¹² | 1.5256 × 10 ⁻¹¹ | 916.50 | 2.95703 × 10 ⁻⁶ | 1.20108 × 10 ⁻⁸ | 2.91329 × 10 ⁻⁸ |
| 278.10 | 8.26154 × 10 ⁻⁸ | 1.57715 × 10 ⁻¹¹ | 4.31927 × 10 ⁻¹¹ | 943.80 | 3.22922 × 10 ⁻⁶ | 1.29326 × 10 ⁻⁸ | 3.30911 × 10 ⁻⁸ |
| 317.70 | 1.23171 × 10 ⁻⁷ | 3.68719 × 10 ⁻¹¹ | 9.71916 × 10 ⁻¹¹ | 970.80 | 3.51437 × 10 ⁻⁶ | 1.53246 × 10 ⁻⁸ | 3.82595 × 10 ⁻⁸ |
| 355.80 | 1.73012 × 10 ⁻⁷ | 6.46005 × 10 ⁻¹¹ | 1.89057 × 10 ⁻¹⁰ | 997.60 | 3.81353 × 10 ⁻⁶ | 1.32687 × 10 ⁻⁸ | 4.344 × 10 ⁻⁸ |
| 392.60 | 2.32439 × 10 ⁻⁷ | 1.29191 × 10 ⁻¹⁰ | 3.5246 × 10 ⁻¹⁰ | | | | |

A.2 SKEW-T LOG-P DIAGRAM FROM STAVANGER

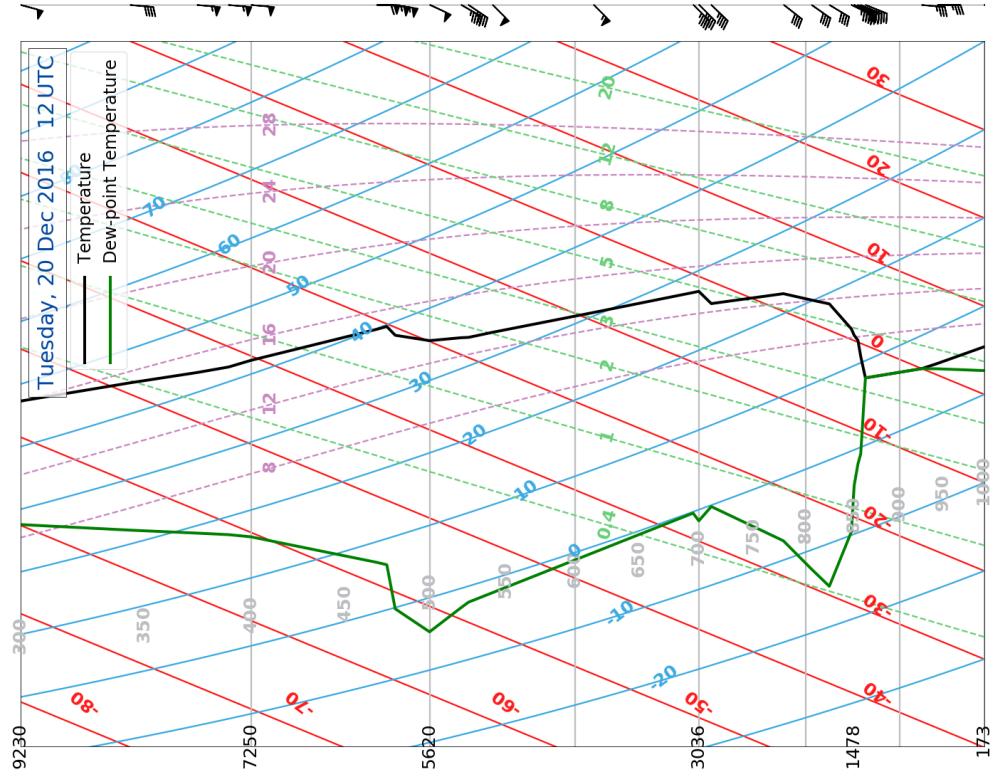
The Skew-T log-P diagram shows the observed vertical temperature and dew-point temperature at Stavanger. The data are taken from ? and processed in Python.

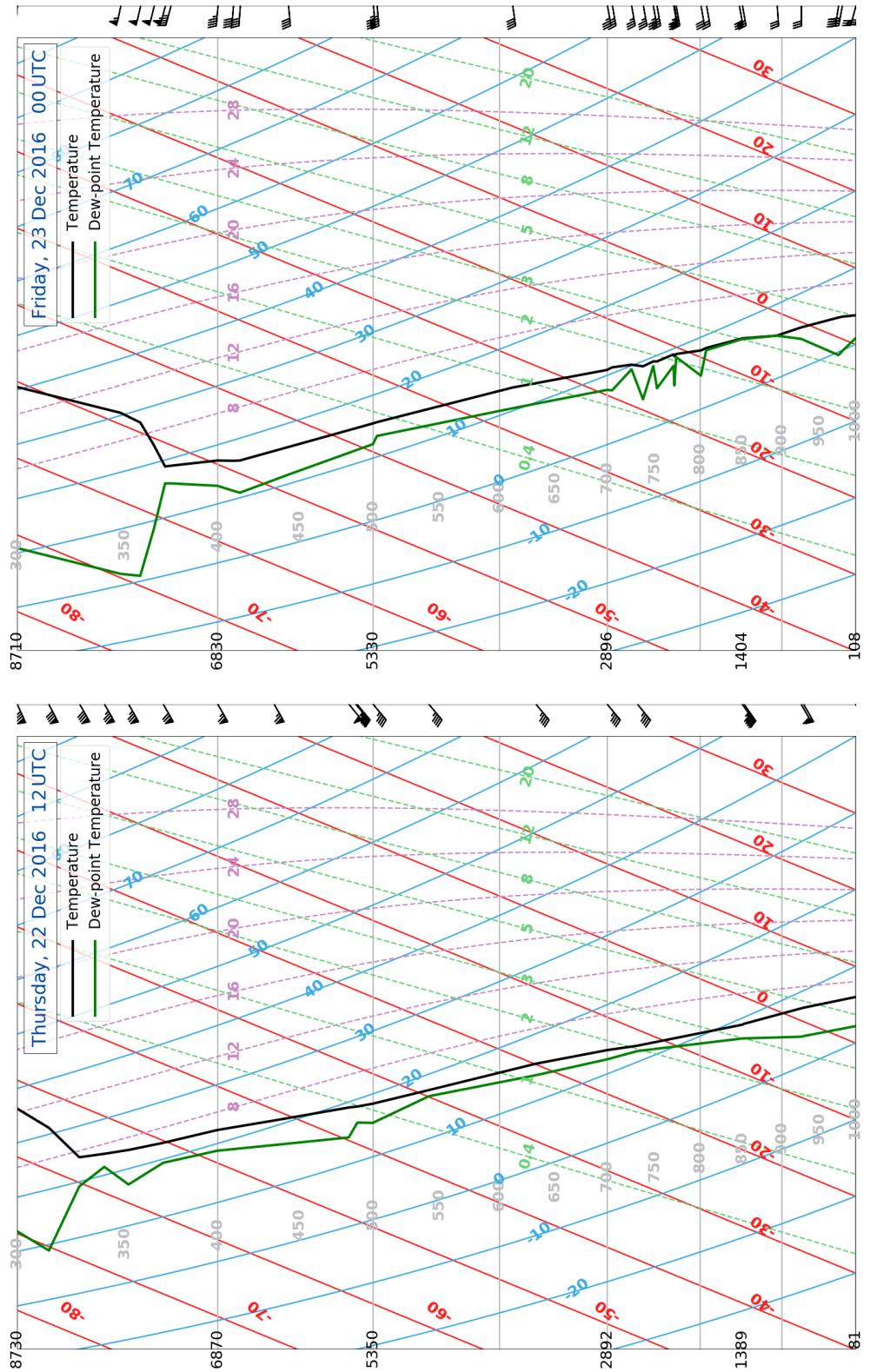
Isobars are grey lines, every 50 hPa, dry adiabats are blue (labelled in $^{\circ}\text{C}$), isotherms are red [$^{\circ}\text{C}$], water vapour mixing ratios are green, dashed in $[\text{g kg}^{-1}]$, and moist adiabats are dashed, purple lines (labelled in [$^{\circ}\text{C}$]).

(b)



(a)





(d)

(c)

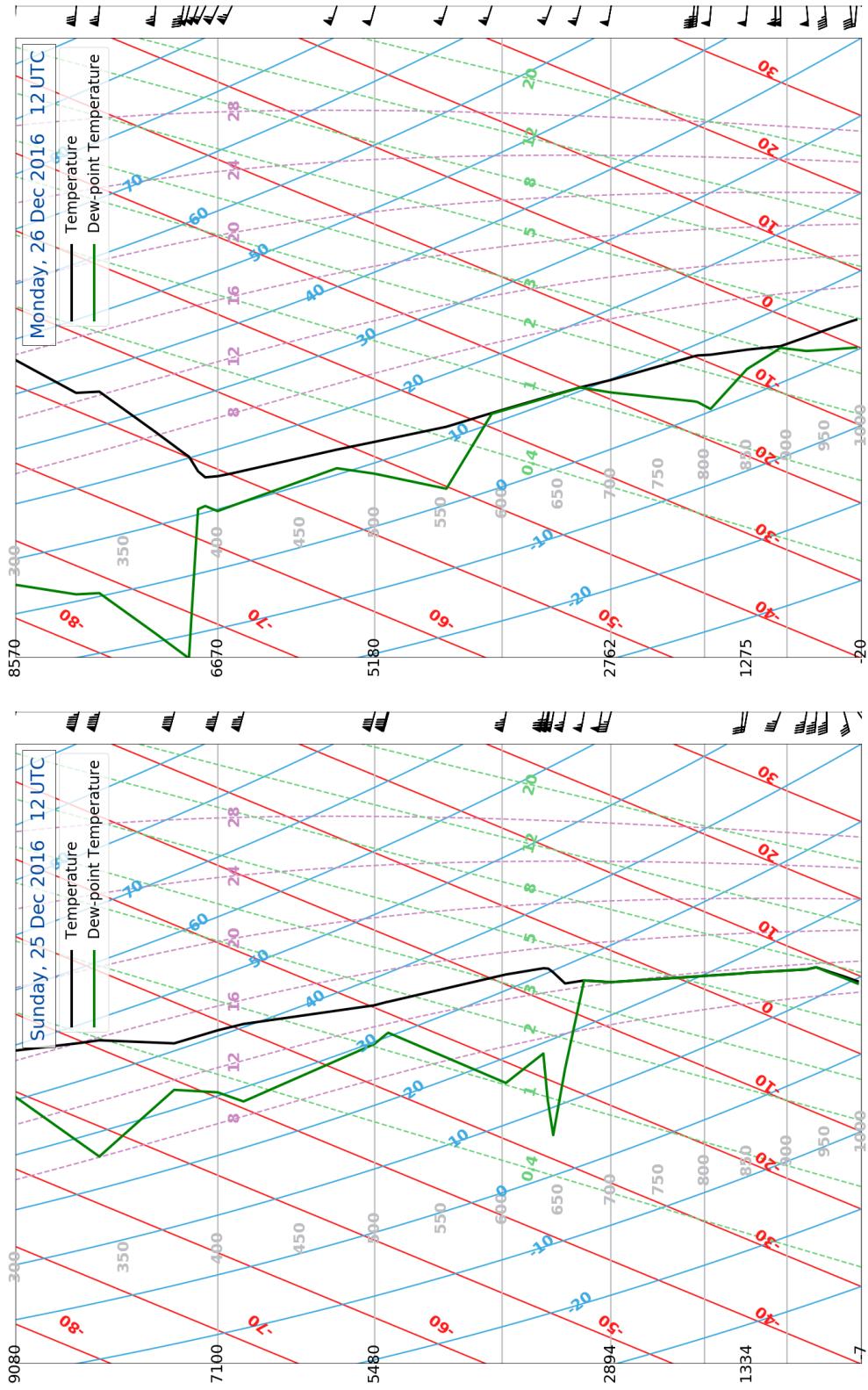


Figure A.2.1: Vertical profiles of atmospheric temperature (black) and dew-point temperature (green) during 20 December 2016 to 26 December 2016. Vertical Profiles from 24 December 2016 are missing at the webpage <http://weather.uwyo.edu/upperair/sounding.html>