Practical aspect for your prediction runs / project

Realtime forecast schedule

Figure 20.12

Hypothetical forecast schedule, for a 00 UTC initialization. A: wait for weather observations to arrive.

- B: data assimilation to produce the analysis (ICs).
- C: coarse-mesh forecast.
- D: fine-mesh forecast, initialized from 00 UTC.
- E: fine-mesh forecast initialized from coarse forecast at 12 h.
- F: postprocessing and creation of products (e.g., weather maps).

Fig. 20.12 shows a hypothetical forecast schedule, for a weather forecast initialized from 00 UTC synoptic observations. First, it takes a few hours (timeline A in Fig. 20.12) for all the data to be communicated from around the world to the weather forecast center (WFC). This step includes quality control, and rejection of suspected bad data.

Next, the data assimilation programs run for a few hours (B) to create a gridded analysis field. This is the optimum initial condition for the NWP model. At this point, we are ready to start making the forecast, but the initial conditions are already 6 h old compared to the present weather.

So the first part of forecast (C) is spent trying to catch up to "present". This wasted initial forecast period is not lamented, because startup problems associated with the still-slightly-imbalanced initial conditions yield preliminary results that should be discarded anyway. Forecasts that occur AFTER the weather has already happened are known as hindcasts, as shown by the shaded area in Fig. 20.12.

The computer continues advancing the forecast (C) by taking small time steps. As the NWP forecast reaches key times, such as 6 , 12, 18, and 24 (=00) UTC, the forecast fields are saved for post-processing and display (F). Lead time is how much the forecast is ahead of real time. For example, for coarse-mesh model (C), weather-map products (F) that are produced for a **valid time** of 18 UTC appear with a lead time of about 8 h before 18 UTC actually happens, in this hypothetical illustration.

NWP meteorologists always have the need for speed. Faster computers allow most phases of the forecast process to run faster, allowing finer-resolution forecasts over larger domains with more accuracy and greater lead time. Speed-up can also be

A: Getting observation data

B: Data Assimilation run

C: Fast / low computational cost model run

D/E: Slow / costly model run

Mesh Design tips

Grid Spacing & smallest resolved features

INFO • Resolution vs. Grid Spacing

Theoretically, the smallest horizontal wavelength you can resolve with data at discrete grid points is $2\Delta X$. However, the finite-difference equations that are used to describe advection and other dynamics in NWP models are unable to handle $2.4X$ waves Namely, these waves either do not advect at all (Fig. 20.11d), or they are numerically unstable.

To avoid such unphysical behavior, small wavelength waves are numerically filtered out of the model. As a result, the smallest waves that are usually retained in NWP models are about 5 to 7.4X.

Hence, the actual **resolution** (i.e., the smallest weather features that can be modeled) are about 7 times the **grid spacing**. Stated another way, if you know the size of the smallest weather system or terrain-related flow that you want to be able to forecast, then you need to design your NWP model with horizontal grid spacing ΔX smaller than 1/7 of that size.

Sample Application

What grid size, domain size, number of grid points, and time steps would you use for a numerical model of a hurricane, and how many computations would be needed to make a 3-day forecast? How fast should your computer be? [Hint: Use info from the Hurricane chapter.]

Assume tropical thunderstorms are about 14 km in diameter

Stull 2017 Practical Meteorology https://www.eoas.ubc.ca/books/Practical_Meteorology/ Chapter 20 Numerical Weather Prediction (NWP)

Figure 16.1 Visible-spectrum satellite picture of Hurricane Katrina over the Gulf of Mexico, taken 28 Aug 2005 at 1545 UTC. (GOES image courtesy of US DOC/NOAA.)

Mesh design

Finest resolution part

- for areas that the thunderstorm / eyewall would possibly pass
	- during the whole simulation
- Grid spacing: <feature length scale> / 7

Surrounding Environment

- Affecting TC track
	- Subtropical high ?
	- \circ Monsoon ?
	- Another TC ?
- Appropriate domain size
- Appropriate grid spacing to resolve feature needed

Assume the smallest feature you want to resolve is a thunderstorm in the eyewall. If tropical thunderstorms are about 14 km in diameter, then you would want ΔX = (14 km)/7 = 2 km to horizontally resolve it.

Hurricanes can be 300 km in diameter. To model the whole hurricane and a bit of its surrounding environment, you might want a horizontal domain of 500 km by 500 km. This works out to $(500 \text{ km} / 2 \text{ km}) =$ 250 grid points in each of the x and y directions, giving $(250)^2$ = 62,500 grid points in the horizontal. If you want a model with 50 vertical levels, then you need $(50) \cdot (62,500) = 3,125,000$ grid points total.

Other settings in Mesh Specification

Estimation details

- Preview of transition
- Number of cells estimation is very preliminary
	- If you turned on "Boost orography" or "Boost coastline", the resulting number of cells may be a lot more than estimated #cells (Note the "≥" symbol)

Max. Resolution Gradient:

- If the target of simulation is to predict / analyse rain.
- Smooth transition of resolution is recommended.
- The default $1/12$ may be too steep (while economical).
- Recommend 1/36 or even smoother, increasing the cost.

Aspects in Real **Simulations**

Initial Value Problem

Needs Initial Condition (IC) data given only.

Evolution of states is done by the model solely.

Atmospheric state Land surface state (soil moisture)

Time-integration

Initial Condition (whole 3-D domain) given

Boundary value problem

From Wikipedia, the free encyclopedia

In mathematics, in the field of differential equations, a boundary value problem is a differential equation together with a set of additional constraints, called the boundary conditions.^[1] A solution to a boundary value problem is a solution to the differential equation which also satisfies the boundary conditions.

Boundary value problems arise in several branches of physics as any physical differential equation will have them.

Lateral boundary condition (LBC) data must be given periodically (e.g. hourly, or 3 hourly) for the lateral boundary (2-D: horizontal boundary perimeter x vertical dimension)

time |
IC LBC LBC LBC LBC time

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Initial Condition

Realtime forecasting for the future

US NCEP **GFS**

- The US NCEP GFS forecast data product.
- Near real-time, some hours of delay
- With GDAS (Global Data Assimilation System)

Should have Data assimilation done - got all available information of observation data

Data source - For real-data Initial Condition

C GFS ?

License: Open Data. There are no restrictions on the use of this data.

 O FNL (2)

O ERA5 2

Re-simulate historic event

ERA5

- The ECMWF ERA5 reanalysis dataset.
- The data has a 5 days delay. ERA5 data since 1979 shall be available.

FNL

- The US NCEP FNL (Final) Operational Global Analysis dataset.
- The data usually has one or two days of delay. FNL data since 2015-07-08 shall be available.

State Imbalance & Spin-up

Imported Initial Condition or Data assimilation introduced imbalanced state.

Analogy:

An extra mass of water is added to a grid cell "suddenly" (due to inference from observation data)

Then this results in spurious wave until it is dissipated.

Figure 20.13

Demonstration of a dynamic system becoming balanced. (a) Balanced initial state of a pond of water (shaded grey), with no waves and no currents. (b) Extra water added in center of pond, causing the water-mass distribution to not be in equilibrium with the waves and currents. (c) Wave generation as the pond adjusts itself toward a new balanced state. (d) Final balanced state with slightly higher water everywhere, but no waves and no currents.

The transient waves and currents are an artifact of the poor initial conditions in the model, and are not representative of the true flow in the real pond. Hence, the forecast results are not to be trusted during the first few minutes of the forecast period while the model is adjusting itself to a balanced state.

Numerical forecasts of the atmosphere have the same problem, but on a longer time scale than a pond. Namely, the first 0.5 to 3 hours of a weather forecast are relatively useless while the model adjusts to imbalances in the initial conditions (see the Data Assimilation section). During this startup period, simulated atmospheric waves are bouncing around in the model, both vertically and horizontally.

After the first 3 to 12 h of forecast, the dynamics are fairly well balanced, and give essentially the same forecast as if the fields were balanced from the start. However, spurious waves in the model might also cause unjustified rejection of good data during data assimilation (see next subsection).

Throw away the data in the spin-up period

Usually, the first \sim 12 hours of mesoscale (around or >10km grid spacing) atmospheric simulation is regarded as the spin-up.

Don't regard it as a valid forecast. Skip the spin-up period in forecast data dissemination and analysis.

Also, the erroneous waves can generate erroneous clouds that cause erroneous precipitation, etc. The net result could be an unrealistic loss of water from the model that could reduce the chance of future cloud formation and precipitation. Change of water content is just one of many **irreversible pro**cesses that can permanently harm the forecast.

In summary, initialization problems cause a transient period of poor forecast quality, and can permanently degrade longer-term forecast skill or cause rejection of good data. Hence, data-assimilation methods to reduce startup imbalances, such as described next, are highly desirable.

Simulation Outputs

Specify what simulation outputs to write to files

Prognostic vs Diagnostic variables

Prognostic:

- Rate of change formulated in the model.
- Need to simulate its time evolution.

Diagnostics:

Can be calculated by other prognostic variables.

Sub-Grid Scale Physics Parameterization

Physics Parameterization applies different length scales

… even finer

A number of important physical processes are not resolved by the grid.

Needs various formulations to calculate parameterized source and sink terms for the PDEs. Grey zone for Cumulus parameterization

Grey zone for Gravity Wave Drag

Mesoscale (coarser than $\sim 10 \text{km}$) \sim Convection-permitting (finer than $\sim 4 \text{km}$)

Grey zone for Boundary Layer turbulence

Many physical processes are resolved by the grid.

Turn off those parameterization

Variable-resolution mesh needs Scale-aware parameterization schemes

Challenging - frontier research underway. Existing schemes are not perfect, a long way to improve.

Brief introduction of NWP Physics

Available choices in WRF

https://www2.mmm.ucar.edu/wrf/users/physics/phys_references.html (Full list and reference to paper)

New Tiedtke

Scheme

option 16

5923-5941.
| doi:10.1175/JCLI-D-16-0597.1
| PDF

Planetary Boundary Laver (PBL) Physics Options (b) ph physics)

Accumulation of science community's contributions - very numerous.

doi:10.2151/jmsj.80.99 **PDF**

Hong, Song-You, Jimy Dudhia, and Shu-Hua Chen, 2004: A revised

NOAA, cited 2001: National Oceanic and Atmospheric Administration Changes to the NCEP Meso Eta Analysis and Forecast System: Increase in

clouds and precipitation. Mon. Wea. Rev., 132, 103-120.

doi:10.1175/1520-0493(2004)132<0103:ARATIM>2.0.CO:2

option 5 resolution, new cloud microphysics, modified precipitation assimilation.

modified 3DVAR analysis. [Available online at

approach to ice microphysical processes for the bulk parameterization of

option 2

options

384

PDF

Scheme

WRF Single-

moment 3-

class and 5-

class Schemes

Eta (Ferrier)

Scheme

Zhang, C. and Y. Wang, 2017: Projected Future Changes of Tropical Cyclone Activity over the Western North and South

Pacific in a 20-km-Mesh Regional Climate Model, J. Climate, 30,

Available choices in WRF

https://www2.mmm.ucar.edu/wrf/users/docs/user_guide_V3/user_guide_V3.8/users_guide_chap5.htm#Phys (Short description of models and options)

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cumulus potential scheme. The scheme also computes the cumulus cloud fraction based on the time-scale relevant for shallow cumuli. (Berg et al. 2013.) New in V3.8.

Available choices in MPAS-A / CPAS

Table 6.3: Possible options for individual physics parameterizations. Namelist variables should be added to the &physics namelist record.

Physics Suite

Mesoscale reference **Convection permitting**

Do the grid cells resolve cumulus?

Small grid spacing (<3km)

Convection resolving / permitting. Updraft and downdraft may be simulated.

Precipitation calculated by Microphysics alone. Turn off convection parameterization

Large grid spacing (>10km)

Grid columns completely contain convective clouds. Convection not resolved, needs parameterization.

> Precipitation calculated by Microphysics + convection parameterization

Interactions of Parameterizations

Moisture distribution may also be affected by other modules.

Cloud affects radiation.

[https://homepages.see.leeds.ac.uk/~lecag/wiser/sampl](https://homepages.see.leeds.ac.uk/~lecag/wiser/sample_wiser_files.dir/Physics_Dudhia.ppt.pdf) [e_wiser_files.dir/Physics_Dudhia.ppt.pdf](https://homepages.see.leeds.ac.uk/~lecag/wiser/sample_wiser_files.dir/Physics_Dudhia.ppt.pdf) Overview of WRF Physics

Microphysics

A emulation of the processes by which moisture is removed from the air, based on other thermodynamic and kinematic fields represented within a model.

Clouds can be resolved but hydrometeors are subgrid

Parameterize micro-scale phenomena like:

- Moisture saturation
- Droplet formation & growth / evaporation
- Raindrop / snow / graupel / hail falling

Need to represent hydrometeor particle size distribution.

<var_array name="scalars" type="real" dimensions="nVertLevels nCells Time"> <var name="qv" array_group="moist" units="kg kg^{-1}" description="Water vapor mixing ratio"/>

> <var name="gc" array group="moist" units="kg kg^{-1}" description="Cloud water mixing ratio" packages="bl_mynn_in;bl_ysu_in;cu_tiedtke_in;mp_kessler_in;mp_thompson_in;mp_wsm6_in"/>

<var name="qr" array group="moist" units="kg kg^{-1}" description="Rain water mixing ratio" packages="mp_kessler_in;mp_thompson_in;mp_wsm6_in"/>

<var name="qi" array_group="moist" units="kg kg^{-1}" description="Ice mixing ratio" packages="bl mynn in; bl ysu in; cu tiedtke in; mp thompson in; mp wsm6 in"/>

<var name="qs" array_group="moist" units="kg kg^{-1}" description="Snow mixing ratio" packages="mp thompson in; mp wsm6 in"/>

<var name="gg" array group="moist" units="kg kg^{-1}" description="Graupel mixing ratio" packages="mp_thompson_in;mp_wsm6_in"/>

<var name="ni" array group="number" units="nb kg^{-1}" description="Cloud ice number concentration" packages="bl_mynn_in;mp_thompson_in"/>

<var name="nr" array_group="number" units="nb kg^{-1}" description="Rain number concentration" packages="mp_thompson_in"/>

</var array>

û

Simple to complex cloud models

Physics only ? Physics + Chemistry? Water only? Aerosol?

Simple: "Single moment":

moisture distribution described by mass ratio only.

Unit: kg/kg weight of moisture / weight of dry air

Aerosol-Cloud Interactions in grid-scale clouds

https://ruc.noaa.gov/wrf/wrf-chem/wrf_tutorial_2018/AerosolInteractions.pdf WRF-Chem tutorial

More complicated cloud model

"Double moment"

moisture distribution described by mass ratio $\{q\bullet\}$ and number concentration {n•}.

Unit: number of droplets / weight of dry air

Convection Parameterization

Cumulus Schemes

- Use for grid columns that completely contain convective clouds
- Re-distribute air in column to account for vertical convective fluxes
	- Updrafts take boundary layer air upwards
	- Downdrafts take mid-level air downwards
- Schemes have to determine
	- When to trigger a convective column
	- How fast to make the convection act

Triggers

- Clouds only activate in columns that meet certain criteria
	- Presence of some convective available potential energy (CAPE) in sounding
	- Not too much convective inhibition (CIN) in sounding (cap strength)
	- Minimum cloud depth from parcel ascent

https://homepages.see.leeds.ac.uk/~lecag/wiser/sample_wiser_files.dir/Physics_Dudhia.ppt.pdf Overview of WRF Physics, Dudhia NCAR

Convection Parameterization Con't

Closures

- Closure determine cloud strength (mass-flux) based on various methods
	- Clouds remove CAPE over time
		- Specified CAPE-removal time scale (KF, Tiedtke, ZM, BMJ)
		- Quasi-equilibrium (Arakawa-Schubert) with large-scale destabilization d(CAPE)/dt (SAS, NSAS)
		- Column moisture convergence
		- Low-level large-scale ascent (mass convergence)

Ensemble methods

- GF, G3 and GD use ensemble of triggers and closures possibly with with varying parameters (effectively up to 144 members)
- Take mean of ensemble to feed back to model
- In principle, can be tuned to emphasize various members under different conditions

Turbulence (Richardson 1922)

Eddies of different sizes.

Lewis Fry Richardson

Numerical approaches

- Direct numerical simulation (DNS)
- Large eddy simulation (LES) first explored by Deardorff (1970 $\tilde{u}_i(\mathbf{x},t) = G_{\Delta} * u_i = \int G_{\Delta}(\mathbf{x}-\mathbf{x}') u_i(\mathbf{x}') d^3\mathbf{x}'$
- Reynolds-averaged Navier-Stokes equation (RANS)

 $\widehat{\overline{u}}(\omega) = \widehat{G}(\omega) \widehat{u}(\omega)$ $\overline{\overline{u}} \neq \overline{u}, \quad \overline{u\overline{v}} \neq \overline{u} \overline{v}$

DNS. LES and RANS

Reynolds-Averaged Navier-Stokes equations (RANS)

Reynolds decomposition of the flow variables into mean and perturbation parts, Reynolds, Osborne (1895)

The flow variables can be u, v, w, T and \theta,

(averaging over a grid volume and period of time)

$$
u_i(x_k, t) = U_i(x_k) + u(x_k, t)
$$

$$
U_i(x_k) = \lim_{T \to \infty} \frac{1}{T} \int_0^T u(x_k, t) dt
$$

 (1) (1) (1) (1) (1)

RANS

averaging over a grid volume and period of time.

$$
\overline{\phi'} = 0
$$
\n
$$
\overline{\phi'} = \overline{\psi'} = 0, \quad \overline{\phi\psi} = \overline{\phi}\overline{\psi} + \overline{\phi'\psi'}, \quad \overline{\phi\phi'} = \overline{\psi\psi'} = \overline{\phi\psi'} = \overline{\psi\phi'} = 0
$$
\n
$$
\overline{\phi^2} = \overline{\phi^2} + \overline{\phi'^2}, \quad \overline{\frac{\partial\phi}{\partial t}} = \frac{\partial\overline{\phi}}{\partial t}, \quad \overline{\frac{\partial\phi}{\partial x_i}} = \frac{\partial\overline{\phi}}{\partial x_i}
$$
\n
$$
(2.5)
$$

$$
\overline{u}_i = \frac{1}{T} \int_t^{t+T} u_i \ dt
$$

 $\frac{1}{T} \int_{t}^{t+T} u'_i \; dt \; = \; 0$

RANS

$$
\frac{\partial u_i}{\partial t} + \frac{\partial}{\partial x_j} (u_i u_j) = -\frac{\partial p}{\partial x_i} + \nu \frac{\partial^2 u_i}{\partial x_j \partial x_j}
$$

$$
\frac{\partial \overline{u}_i}{\partial t} + \frac{\partial}{\partial x_j} (\overline{u_i u_j}) = -\frac{\partial \overline{p}}{\partial x_i} + \nu \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j}
$$

$$
u = \overline{u} + u'
$$

$$
w = \overline{w} + w'
$$

$$
p = \overline{p} + p'
$$

$$
\overline{u_i u_j} = (\overline{u}_i + u'_i)(\overline{u}_j + u'_j) = \overline{u}_i \overline{u}_j + \overline{u}_i u'_j + \overline{u'_i u_j} + \overline{u'_i u'_j} = \overline{u}_i \overline{u}_j + \overline{u'_i u'_j}
$$

$$
\frac{\partial \overline{u}_i}{\partial t} + \frac{\partial}{\partial x_j} (\overline{u}_i \overline{u}_j) = -\frac{\partial \overline{p}}{\partial x_i} + \nu \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j} - \frac{\partial}{\partial x_j} (\overline{u'_i u'_j})
$$

u', w' are correlated

 $\overline{w'} = 0$
 $\overline{u'} = 0$

but...

 $\overline{w'u'}\neq 0$

Reynolds-stress term

$$
\frac{\partial \overline{u}_i}{\partial t} + \frac{\partial}{\partial x_j} (\overline{u}_i \overline{u}_j) = -\frac{\partial \overline{p}}{\partial x_i} + \nu \frac{\partial^2 \overline{u}_i}{\partial x_j \partial x_j} - \frac{\partial}{\partial x_j} (\overline{u'_i u'_j})
$$

 $\frac{\partial}{\partial z}\left|\nu \frac{\partial \overline{u}}{\partial z}\right|$ -Viscous force (molecular viscosity, diffusion by molecular motions).

-The complete form is Reynolds-stress tensor, is symmetric, the diagonal components are normal stress and off-diagonal components are shear stress.

$$
R_{ij} = -\rho \left(\begin{array}{ccc} \overline{u'u'} & \overline{u'v'} & \overline{u'w'} \\ \overline{v'u'} & \overline{v'v'} & \overline{v'w'} \\ \overline{w'u'} & \overline{w'v'} & \overline{w'w'} \end{array} \right)
$$

Theories at a glance (details skipped)

Eddy viscosity coefficient
 $\overline{(u'w')} = -K_m \frac{\partial \overline{u}}{\partial z}$ Prandtl Mixing Length Theory $K_m = l_v^2 \left| \frac{\partial \bar{u}}{\partial z} \right|$

Friction velocity on surface layer

$$
u_*^2 \equiv \left| \overline{(u'w')}_{s} \right|
$$

 $\overline{u} = \frac{u_*}{\kappa} \ln \left(\frac{z - d}{z_0} \right)$

Monin-Obukhov Length

$$
L\equiv\frac{-u_*^3\bar{\theta}_v}{\kappa g\overline{\left(w'\theta_v'\right)_s}}
$$

Modification of log wind profile

$$
\frac{d\overline{u}}{dz} = \frac{u_*}{\kappa z} \phi_m
$$

$$
d\overline{u} = \frac{u_*}{\kappa} \left[\frac{dz}{z} - (1 - \phi_m) \frac{dz/L}{z/L} \right]
$$

$$
\overline{u} = \frac{u_*}{\kappa} \left[\ln \frac{z}{z_0} - \psi_m \right]
$$

$$
u_{10m} = u_a \frac{\ln\left(\frac{10+z_0}{z_0}\right) - \psi_m\left(\frac{10+z_0}{L}\right) + \psi_m\left(\frac{z_0}{L}\right)}{\ln\left(\frac{z+z_0}{z_0}\right) - \psi_m\left(\frac{z+z_0}{L}\right) + \psi_m\left(\frac{z_0}{L}\right)},
$$
\n(24)

$$
\theta_{2m} = \theta_g + (\theta_a - \theta_g)
$$

$$
\times \frac{\ln\left(\frac{2+z_0}{z_0}\right) - \psi_h\left(\frac{2+z_0}{L}\right) + \psi_h\left(\frac{z_0}{L}\right)}{\ln\left(\frac{z+z_0}{z_0}\right) - \psi_h\left(\frac{z+z_0}{L}\right) + \psi_h\left(\frac{z_0}{L}\right)},
$$
(25)

$$
q_{2m} = q_g + (q_a - q_g)
$$

$$
\times \frac{\ln\left(\frac{\rho c_p k u_* 2}{c_s} + \frac{2}{z_l}\right) - \psi_h\left(\frac{2}{L}\right) + \psi_h\left(\frac{z_l}{L}\right)}{\ln\left(\frac{\rho c_p k u_* z}{c_s} + \frac{z}{z_l}\right) - \psi_h\left(\frac{z}{L}\right) + \psi_h\left(\frac{z_l}{L}\right)}.
$$
 (26)

Review surface layer scheme

PBL schemes

Planetary Boundary Layer Scheme

FIG. 33. Time evolution of the computed (basic case) and observed mixed layer height.

Yamada and Mellor (1975)

PBL-Daytime(Afternoon)

- Strong surface heating, the virtual potential temperature will decrease with height near the surface ground. (close to surface layer)
- Moreover, the convective turbulence or eddies mixes efficiently and it is in the mixed layer.
- Vertical profiles of virtual potential temperature, vapor mixing ratio and horizontal momentum. (time-averaged is better)

Contrast day vs. night

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Modeling PBL -- YSU scheme Yonsei University Scheme

- 1. K-profile parameterization (KPP) over the depth of PBL
- 2. Another is based on turbulence kinetic energy (TKE)

The KPP was discussed as non-local K -theory [1986] and is supported by large eddy simulation. In surface layer scheme, K_m is defined as,

$$
\overline{(u'w')} = -K_m \frac{\partial \overline{u}}{\partial z}
$$

 $K_m = l_v^2 \left| \frac{\partial \overline{u}}{\partial z} \right|$

and \underline{a} simple KPP formula of K_m is,

$$
K_m = kw_s z \left(1 - \frac{z}{h}\right)^p,
$$

\kappa is von Karman constant(=0.4), z is height from surface, h is PBL depth, p=2 in usual, w s is velocity scale at surface (u* is friction velocity

and \phi_m is stability correction in surface layer)

$$
w_s = (u_*^3 + \phi_m k w_{*b}^3 z/h)^{1/3},
$$

Troen, I., and L. Mahrt, 1986: A simple model of the atmospheric boundary layer sensitivity to surface evaporation. Bound. Layer Meteor., 37, 129–148.

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The typical variation of eddy viscosity K

From observation,

Surface Layer

YSU (diagnostic scheme) imposes this MYJ (prognostic scheme) tries to develop it

$$
K_m = kw_s z \bigg(1 - \frac{z}{h}\bigg)^p,
$$

 $W_s = (u_*^3 + \phi_m k w_{*b}^3 z/h)^{1/3}$,

u* is the surface friction velocity w^{*} is the convective velocity scale on surface b is moist air

$$
\text{moist air, } w_{\ast b} = \left[\left(\frac{g}{\theta_{va}} \right) \left(\overline{w' \theta_v'} \right) \right]^{1/3}.
$$

[H06]

label a is at the lowest model level, label 0 is near surface

 O'Brien, J. (1970), A note on the vertical structure of the eddy exchange coefficient in the PBL,J. Atmos. Sci.,27: 1213–1215.

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z

 h

 $2s₁$

 (1988)

Counter-gradient term

In YSU, the counter-gradient term is applied to temperature, water vapor mixing ratio and momentum. (details on /physics_wrf/module_bl_ysu.F)

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PBL Structure and Heat flux

Warner (2011)

PBL-Daytime(Afternoon)--Example:

92-25km mesh, it may not be finer enough

HK 2020-8-14-06 (UTC+8 = 2 pm) by Physics Suite: mesoscale reference (YSU)

Unstable atmosphere: PBL height (hpbl = 1070.5765 m)

(Unstable atmosphere: PBL height(hpbl = 1070.5765 m)

PBL-Nighttime--Example: HK

HK 2020-8-14-18 (UTC+8 = 2 am) by Physics Suite: mesoscale reference(YSU)

stable atmosphere: PBL height(hpbl = 337.11786 m)

stable atmosphere: PBL height(hpbl = 337.11786 m)

Wind speed with height

Fig. 4.9 Diurnal cycle of wind speed as a function of height measured from a tower in Oklahoma City and averaged over the period June 1966 to May 1967. [Adapted from Crawford and Hudson (1973). Reprinted with permission from the American Meteorological Society.]

How to learn more on Physics Parameterization Schemes

How to learn more about Physics Parameterizations?

The hard ways:

- Original papers:
	- E.g. How are diagnostics u10 calculated
	- "Surface layer model"
- Reference found in (some) source code
	- Some source code has a documentation header section.
	- Not all source code are well-documented

A more digestible way:

- Regular conferences
	- E.g. the P3 microphysics scheme
- Google search for slides and presentations
	- E.g. scale-aware gravity wave drag scheme

Hard: Finding originating papers - CPAS

CPAS: [User Guide - Real Simulation page](https://cpas.earth/userguide/realsimulation)

PBL (PLANETARY BOUNDARY LAYER) SCHEMES

Just a summary of information from WRF for user's convenience.

PBL (PLANETARY BOUNDARY LAYER) SCHEMES

Question: how are t2m, u10, v10 calculated?

Short answer:

- by the "Surface Layer" scheme
- the lowest layer is usually higher than 10m and surely higher than 2m.

https://docs.google.com/spreadsheets/d/1nqeD9wsI1xjlXNMmVYHJJ4lhopKEwCZsqjS2Cbi1Ys4

Hard: Finding originating papers - WRF

WRF: [Physics Reference page](https://www2.mmm.ucar.edu/wrf/users/physics/phys_references.html)

A https://www2.mmm.ucar.edu/wrf/users/physics/phys_references.html#SL

Surface Layer Options (sf sfclay physics) Jimenez, Pedro A., Jimy Dudhia, J. Fidel Gonzalez-Rouco, Jorge Navarro, Juan P. Montavez, and Elena Garcia-Bustamante, 2012: A revised scheme for the WRF surface **Revised MM5 Scheme** option 1 layer formulation. Mon. Wea. Rev., 140, 898-918. doi:10.1175/MWR-D-11-00056.1 **PDF** Monin A. S., and A. M. Obukhov, 1954: Basic laws of turbulent mixing in the surface layer of the atmosphere. Contrib Geophys Inst Acad Sci USSR 151:163-187 (in Russian) **PDF** Janjic, Z. I., 1994: The step-mountain Eta coordinate model: further developments of the convection, viscous sublayer and turbulence closure schemes. Mon. Wea. Rev., 122, 927-945. doi:10.1175/1520-0493(1994)122<0927:TSMECM>2.0.CO:2 PDF **Eta Similarity Scheme** option 2 Janiic, Z. I., 1996: The surface laver in the NCEP Eta Model. Eleventh conference on numerical weather prediction. Norfolk. VA, 19-23 August 1996. Amer Meteor Soc, Boston, MA, pp 354-355. **PDF** Janjic, Z. I., 2002: Nonsingular implementation of the Mellor-Yamada Level 2.5 Scheme in the NCEP Meso model, NCEP Office Note No. 437, 61 pp. **PDF NCEP Global Forecast** option 3 **System Scheme QNSE Scheme** option 4 **MYNN Scheme** option 5

Hard: Finding source code and references

MPAS github

[https://github.com/MPAS-Dev/MPAS-Model/tree/ma](https://github.com/MPAS-Dev/MPAS-Model/tree/master/src/core_atmosphere/physics/physics_wrf) [ster/src/core_atmosphere/physics/physics_wrf](https://github.com/MPAS-Dev/MPAS-Model/tree/master/src/core_atmosphere/physics/physics_wrf)

Code with documentation header

Digestible way: Google search and look for …

(Yearly) NCAR WRF & MPAS workshop

[2019 Agenda](https://www2.mmm.ucar.edu/wrf/users/workshops/WS2019/workshop19agenda.php) [2020 Agenda](https://www2.mmm.ucar.edu/wrf/users/workshops/WS2020/workshop20agenda.php) [2021 Agenda](https://www2.mmm.ucar.edu/wrf/users/workshops/WS2021/workshop21agenda.php) [2022 Agenda](https://www.mmm.ucar.edu/events/workshops/2022/agenda)

Using hierarchical time-stepping to utilize MPAS-A computational resources for customized extreme variable-resolution meshes. Ng, Ka-Ki, Kwan-Shu Tse, Yuk Sing Lui, Wai-Nang Leung, Chi Chiu Cheung, and Sze-Chuan Suen, ClusterTech Limited, Hong Kong presentation

> **Session: Physics Developments / Challenges** First Chair: Wayne Angevine, CIRES/CU Boulder and NOAA/ESRL Second Chair (after break): Lulin Xue, NCAR

Program for the Joint WRF/MPAS Users' Workshop 2020 (Virtual) Location: online Date: 8 - 9 June 2020

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<https://www.youtube.com/watch?v=Kmax80kMddk>

What is "gravity wave breaking" ?

What to pay attention to when experimenting the implemented parameterization scheme?

Ensemble Forecasting

ECMWF's introduction to Ensemble Forecasting

Fact sheet: Ensemble weather forecasting

23 March 2017

What is ensemble weather forecasting?

An ensemble weather forecast is a set of forecasts that present the range of future weather possibilities. Multiple simulations are run, each with a slight variation of its initial conditions and with slightly perturbed weather models. These variations represent the inevitable uncertainty in the initial conditions and approximations in the models. They produce a range of possible weather conditions.

Why is it important to measure the level of uncertainty in a forecast?

The uncertainty associated with every forecast means that different scenarios are possible. and the forecast should reflect that. Single 'deterministic' forecasts can be misleading as they fail to provide this information. Take agriculture as an example: a farmer needs to know the range of possible conditions the crops may experience so that they can be protected. Ensemble forecasts show how big that range is at different forecast times.

What are the advantages of ensemble prediction?

By generating a range of possible outcomes, the method can show how likely different scenarios are in the days ahead, and how long into the future the forecasts are useful. The smaller the range of predicted outcomes, the 'sharper' the forecast is said to be. Good ensemble forecasts are not just as sharp as possible but also reliable. If a reliable forecast says that there is a 70% chance of top temperatures rising above a certain threshold, then in 70% of cases when such a forecast is made temperatures will indeed rise above that threshold.

Is uncertainty in a forecast due to a lack of knowledge?

Yes, to some extent our lack of knowledge does significantly increase uncertainty in the forecast. This is why there is much work going into improving our knowledge of initial conditions and of atmospheric processes that computer models need to mirror. In addition, the atmosphere is a chaotic system. This means that it is sensitively dependent on initial conditions. In a chaotic system, a slight change in the input conditions can lead to a significant change in the output forecast. In a non-chaotic system, small differences in initial conditions only give small differences in output. Hence, it is important in weather forecasting to investigate how sensitive the atmosphere is at any stage to initial conditions. Ensemble forecasting does this by looking at a spread of possible outcomes.

[https://www.ecmwf.int/en/about/media-centre/focus/2017/fact-shee](https://www.ecmwf.int/en/about/media-centre/focus/2017/fact-sheet-ensemble-weather-forecasting) [t-ensemble-weather-forecasting](https://www.ecmwf.int/en/about/media-centre/focus/2017/fact-sheet-ensemble-weather-forecasting)

Chaotic nature of the atmosphere

Well-known "Bufferfly effect".

Prediction results will be different, given minor differences in:

- Initial Condition
- Model and model options

INTRODUCTION 1.

Ensemble Prediction Systems (EPS) are numerical weather prediction (NWP) systems that allow us to estimate the uncertainty in a weather forecast as well as the most likely outcome. Instead of running the NWP model once (a deterministic forecast), the model is run many times from very slightly different initial conditions. Often the model physics is also slightly perturbed, and some ensembles use more than one model within the ensemble (multi-model EPS) or the same model but with different combinations of physical parameterization schemes (multi-physics EPS). Owing

Guidelines on Ensemble Prediction Systems and Forecasting WMO-No. 1091 © World Meteorological Organization, 2012 https://library.wmo.int/doc_num.php?explnum_id=7773

https://en.wikipedia.org/wiki/Ensemble_forecasting

Ensemble forecasting

From Wikipedia, the free encyclopedia

Ensemble forecasting is a method used in or within numerical weather prediction. Instead of making a single forecast of the most likely weather, a set (or ensemble) of forecasts is produced. This set of forecasts aims to give an indication of the range of possible future states of the atmosphere. Ensemble forecasting is a form of Monte Carlo analysis. The multiple simulations are conducted to account for the two usual sources of uncertainty in forecast models: (1) the errors introduced by the use of imperfect initial conditions, amplified by the chaotic nature of the evolution equations of the atmosphere. which is often referred to as sensitive dependence on initial conditions; and (2) errors introduced because of imperfections in the model formulation, such as the approximate mathematical methods to solve the equations. Ideally, the verified future atmospheric state should fall within the predicted ensemble spread, and the amount of spread should be related to the uncertainty (error) of the forecast. In general, this approach can be used to make probabilistic forecasts of any dynamical system, and not just for weather prediction.

Top: Weather Research and Forecasting model simulation of Hurricane Rita tracks. Bottom: The spread of National Hurricane Center multi-model ensemble forecast.

Composing ensemble members for our CPAS project

https://en.wikipedia.org/wiki/Ensemble_forecasting#Multi_model_ensembles

Multi model ensembles [edit]

When many different forecast models are used to try to generate a forecast, the approach is termed multi-model ensemble forecasting. This method of forecasting can improve forecasts when compared to a single model-based approach.^[18]

Different ICs as ensemble members:

- The latest IC at our role-paly weather conference time.
- Available ICs before that latest IC
	- Longer simulation duration needed.
	- Give up information coming from latest observation data.

Top: Weather Research and Forecasting model simulation of Hurricane Rita tracks. Bottom: The spread of Nation multi-model ensemble forecast.

Prof. Robert Fovell (UCLA); Dr. Hui Su (JPL) - http://mls.jpl.nasa.gov/research/hurricanes.php

(Top): WRF model simulation of Hurricane Rita tracks. The model resolution is 30km. The colored field shows the lowest sea-level pressure (SLP) recorded during the last 27 hours of a 54 hour control simulation of Rita using LFO (5 class) microphysics and the Kain-Fritsch (KF) convective scheme. The superposed black line traces the model hurricane, which strikes Houston. Also shown are tracks of minimum SLP for runs using the Kessler (warm rain) scheme, the WSM3 simple ice scheme (with the Betts-Miller-Jancic convective scheme), the Kessler with reduced rain fallspeed, and WSM3 with enhanced ice fallspeed. (Bottom): The spread of NHC multi-model ensemble forecast at 06 UTC, 22 September. Note a similar ensemble spread was obtained from a single model simply by varying the model microphysics and convective schemes. Image from Jonathan Vigh, Colorado State University.

People put ensemble forecasting result together to take a glance

Spaghetti maps 4.1.5

Charts showing a few selected contours of variables (for example, 528, 546 and 564 Dm contours of 500 hPa geopotential height) from all ensemble members can provide a useful image of the predictability of the field. Where all ensemble member contours lie close together the predictability is higher; where they look like spaghetti on a plate, there is less predictability (see Figure 4).

Source: UK Met Office using data from ECMWF, © British Crown Copyright

Figure 4. Ensemble 500 hPa forecast spaghetti charts for 11 February 2001 at 1200 UTC (T + 96 from 7 February 2001 at 1200 UTC)

$4.1.6$ **Postage stamp maps**

A set of small maps showing contoured plots of each ensemble individual member (see Figure 5) allows the forecaster to view the scenarios in each member forecast and assess the possible risks of extreme events. However, this presents a large amount of information that can be difficult to assimilate.

Source: UK Met Office using data from ECMWF, © British Crown Copyright

Figure 5. Postage stamp map for 7 February 2009 at 1200 UTC (850 hPa wetbulb potential temperature, in degrees Celsius; T + 300 from 26 January at 0000 UTC)

Guidelines on Ensemble Prediction Systems and Forecasting WMO-No. 1091 © World Meteorological Organization, 2012 https://library.wmo.int/doc_num.php?explnum_id=7773

More examples

<https://www.encyclopedie-environnement.org/en/air-en/overall-forecast/> http://www.geo-tasks.org/geoss_portfolio/weather_tigge.php

Occurrence probability of extreme 24-hr precipitation Valid: 2011060312UTC +5-6days

Figure 2: The forecast probability of heavy rainfall (exceeding the 95th percentile), based on four TIGGE ensembles, and a multi-centre grand ensemble.

https://cpas.earth/ Summer 2022