

Improving DOE-2's RESYS routine: User Defined Functions to Provide More Accurate Part Load Energy Use and Humidity Predictions

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ABSTRACT

In hourly energy simulations, it is important to properly predict the performance of air conditioning systems over a range of full and part load operating conditions. An important component of these calculations is to properly consider the performance of the cycling air conditioner and how it interacts with the building. This paper presents improved approaches to properly account for the part load performance of residential and light commercial air conditioning systems in DOE-2. First, more accurate correlations are given to predict the degradation of system efficiency at part load conditions. In addition, a user-defined function for RESYS is developed that provides improved predictions of air conditioner sensible and latent capacity at part load conditions. The user function also provides more accurate predictions of space humidity by adding "lumped" moisture capacitance into the calculations. The improved cooling coil model and the addition of moisture capacitance predicts humidity swings that are more representative of the performance observed in real buildings.

Introduction

In hourly energy simulations, it is important to properly predict the performance of air conditioning systems over a range of full and part load operating conditions. An important component of these calculations is to properly consider the performance of the air conditioner and how it interacts with the building.

Energy simulation programs such as DOE-2 use well understood calculation procedures based on "first principles" to determine the heat transfer through the building envelope, or loads. However, the performance of the HVAC equipment is typically determined using empirical functions that predict system energy use and capacity as a function of operating conditions such as outdoor and indoor temperature, humidity, and equipment loading. DOE-2 includes empirical functions in the form of multi-variable polynomials for numerous types of HVAC systems (Buhl et al 1993). These curves typically "extend" the nominal, or design performance of the system to off design conditions. DOE-2 provides default empirical models for these systems that have been developed over the last 20 years. Many of these curves are based on equipment that may no longer reflect products that are available in the market. In other cases the default curves may have been developed based on data over a narrow range of conditions that are no longer appropriate.

Perhaps the most complicated models are required for the simplest system: the residential or small commercial air conditioner. This direct-expansion cooling unit cycles on and off to meet the load based on the space temperature sensed by a thermostat. The relatively complex dynamics of the DX air conditioner and the thermostat are modeled in hourly simulation programs by using simplified empirical functions that predict the

degradation in efficiency (or added energy use) for hours when the unit operates for only part of the time.

Another complicating factor is that the DX air conditioner provides both sensible cooling and moisture removal (i.e., latent capacity). The mix of latent and sensible capacity provided by the unit depends on several factors including ambient temperature, the air flow rate, and the psychrometric conditions entering the evaporator coil¹.

This paper describes methods to improve the predictions of part load air conditioner performance of the residential systems (RESYS) routine in DOE-2. The improvements and features of the model include:

1. More realistic predictions of part load efficiency degradation and its impact on air conditioner energy use.
2. Better predictions of air conditioner latent and sensible capacity over a range of operating conditions.
3. The addition of moisture capacitance into DOE-2's zone moisture balance calculations to improve the prediction of time-varying space humidity levels.

The following section addresses the issue of part load efficiency degradation for DX air conditioners and suggests improved FEIR-PLR functions for use in DOE-2. Then a more rational DX air conditioner model is described and the method for integrating the model into the DOE-2 RESYS routine as a user defined function is presented.

Part Load Efficiency Degradation for Cycling Equipment

It is generally convenient to express part load effects in terms of degradation of efficiency under part load. Parken et al (1977) at NIST (formerly the National Bureau of Standards) referred to the normalized efficiency degradation as the part load factor, or PLF.

$$PLF = \frac{PartLoadEfficiency}{SteadyStateEfficiency} \quad (1)$$

This nomenclature is also used in the SEER test procedures (DOE 1979). PLF of a cycling HVAC system depends on:

1. the response of the cooling system at startup (usually defined by a time constant or dead time),
2. the cycling rate of the equipment (usually defined by thermostat characteristics and to a lesser extent the building thermal mass).

Parken and his co-workers at NIST were the first to recognize that these two factors could be combined to form a part load correlation. They used this concept to develop the part load degradation coefficient (C_d) used today in the SEER rating procedure (equation 2). They verified the concept with both laboratory and field data (Parken et al 1985). PLR, the part load ratio, is defined as the ratio of the hourly load and available capacity.

¹ In addition, the latent and sensible mix of a cycling air conditioner with constant fan operation is also a function of the runtime fraction. A method to consider this latent degradation effect is given by Henderson and Rengarajan (1996).

$$PLF = 1 - C_d(1 - PLR) \quad (2)$$

Henderson and Rengarajan (1996) showed that the theoretical part load efficiency curve could be summarized in the form of Equation 3. Successive substitution is required to solve for PLF as a function of PLR and the parameters N_{max} and τ .

$$PLF_{i+1} = 1 - \alpha(1 - e^{-1/\alpha}) \quad (3a)$$

where :

$$\alpha = 4\tau N_{max} \left(1 - \frac{PLR}{PLF_i}\right) \quad (3b)$$

and:

N_{max} = Maximum cycling rate of the cooling system / thermostat (cycles/h)
 τ = Time constant of air conditioner cooling capacity (hrs)

Figure 1 shows that the theoretical function (eqn. 3) closely matches the linear C_d model (eqn. 2). The default value of C_d in the SEER rating procedure is 0.25. This value corresponds to time constant (τ or tau) of 76 seconds for the air conditioner capacity at startup, and a maximum thermostat cycling rate (N_{max}) of 3.125 cycles per hour. For systems with less degradation (i.e., smaller C_d or τ) the linear curve is even more closely matched to the theoretical model.

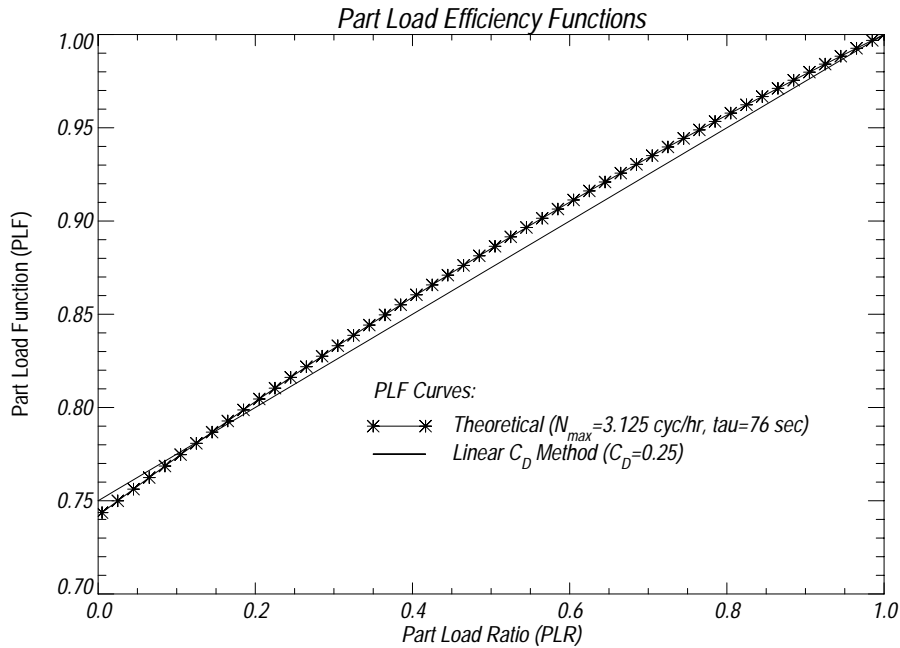


Figure 1. Comparing Theoretical and Linear Part Load Models

Modern air conditioner and heat pump systems typically have time constant of 40 to 60 seconds. As a result, values of C_d measured for typical systems tested according Tests C and D of the SEER procedure are in the range of 0.1 to 0.2. Thermostat cycling rates are also generally lower than the default value of 3.125 assumed in the rating procedure. An average

cycling rate of 30 Florida homes was found to be 2.5 cycles per hour (Henderson et al 1991). This further reduces the effective value of C_d .

For systems that use power when the system is off, it is sometimes desirable to include the impact of this power into the part load degradation curve². Off cycle power can include crankcase heaters, controls, or other parasitic uses.

Bonne et al (1980) and Miller and Jaster (1985) both showed that, when off-cycle power consumption is considered, the part load efficiency curve is modified as given by equation (4) and PLF' is the resulting factor considering off-cycle power (where PLF is from equation 2 or 3). The off-cycle power use is expressed as a fraction of on-cycle power use (pr).

$$PLF' = \frac{PLR}{\left(\frac{PLR}{PLF} + \left(1 - \frac{PLR}{PLF}\right) pr \right)} \quad (4)$$

Figure 2 shows that considering even small amounts of off-cycle power has a dramatic impact on efficiency at low load conditions. A value of 0.01 for pr corresponds to about 40 Watts of off-cycle power for a typical 3 ton AC system, while 0.03 corresponds to 120 Watts.

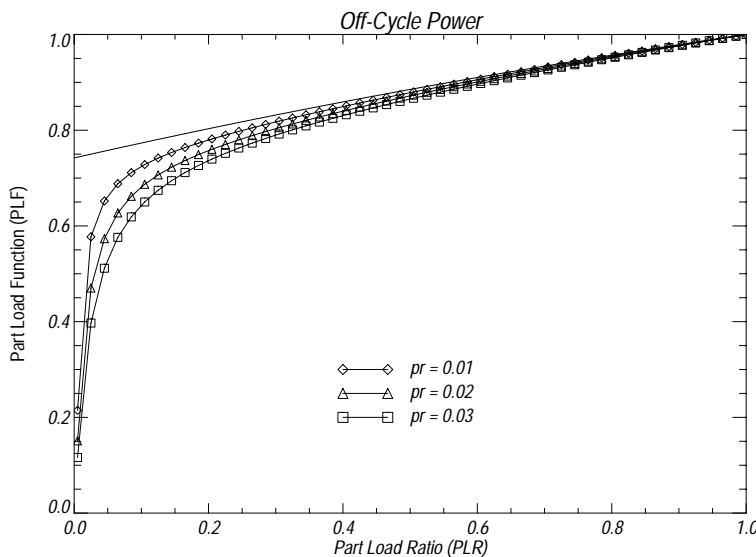


Figure 2. The Impact of Off-cycle Power on PLF

DOE-2's Approach to Part Load Degradation

DOE-2 includes several correlation curves that predict the energy use of systems under part load conditions. Because DOE-2 was first developed for larger buildings that typically use chillers, the efficiency degradation is expressed in terms of the Energy Input Ratio (EIR). EIR is the inverse of efficiency, or COP, and is the dimensionless equivalent of kW per ton. The part load correlations for DOE-2 are polynomials that predict EIR as

² The other way to include this power is to use the equipment runtime fraction and account for that power in a separate calculation.

function of PLR. Then the EIR-FPLR factor is multiplied by the nominal power of the machine to find its part load energy use. The EIR-FPLR functions can be rearranged to find PLF with equation 5.

$$PLF = \frac{PLR}{EIR - FPLR} \quad (5)$$

Figure 3 shows the DOE-2 default curves for part load performance of various types of equipment plotted in the PLF form. Table 1 lists the DOE-2 SYSTEMS and PLANT equipment that use these curves. Due to the inverted form of the equation, all five curves show the following behavior:

- the part load efficiency goes to zero as PLR approaches zero,
- the slope of the curves is strongly a function of loading (i.e., PLR)

The curve for residential cooling shows the most part load degradation, followed by the boiler and heat pump heating curves. The furnace and PSZ cooling curves show the least amount of part load degradation. Even though these curves are a simple linear model of EIR as function of PLR, when inverted using equation 5 the functions predict a significant amount of degradation that was not readily apparent in the original EIR form.

Table 1. Default Part Load Curves in DOE-2

Description	Curve Name	Curve No	DOE-2 Systems
Residential Cooling	COOL-EIR-FPLR	16,17,20	RESYS,PTAC,HP
Commercial Cooling	COOL-EIR-FPLR	18, 128	PSZ,PMZS,PVAVS PVVT
HP Heating	HEAT-EIR-FPLR	61,62,65, 75,116	RESYS, PSZ, PTAC, PVAVS, HP, WTR-CC, PVVT
Furnace	FURNACE-HIR-FPLR	111	any fuel-fired furnace
Boiler	BOILER-HIR-FPLR	BLRHIR2	HP (WLHP system) HW and Steam Boiler Plants

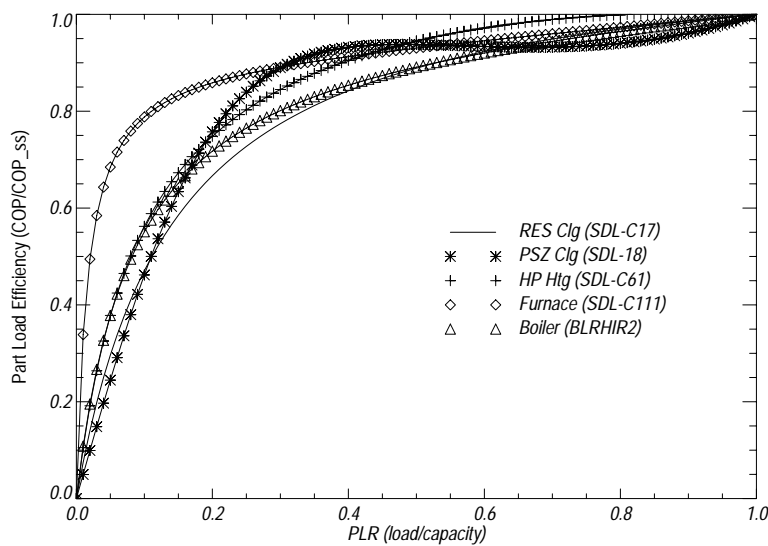


Figure 3. Default Curves for DOE-2 Presented in the PLF vs. PLR Form

Recommended Curves for Residential and Light Commercial Cooling Systems

Figure 4 shows the range of part load performance that might be expected for a residential cooling system. Table 2 lists the corresponding EIR coefficients. The two plots show curves that include the impact with and without off-cycle power. The default RESYS curve (without off-cycle power adjustments) is also shown on each plot for reference.

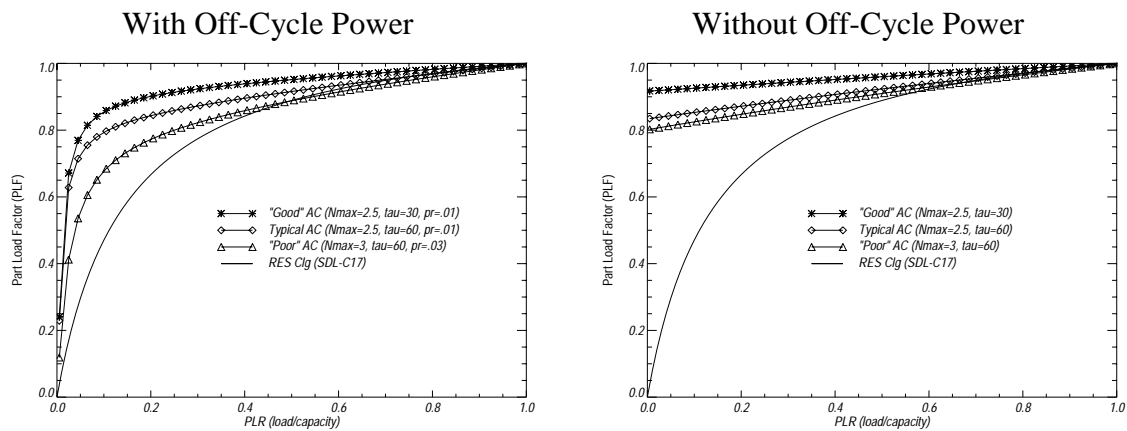


Figure 4. Recommended Parameters for a Typical AC System Compared to Default DOE-2 Curve (without off-cycle power)

Table 2. EIR Coefficients For “Typical AC” in Figure 4

	Coefficients for $EIR-FPLR = a + b \cdot PLR + c \cdot PLR^2 + d \cdot PLR^3$			
	a	b	c	d
Curves with Off-Cycle Power				
“Typical AC” ($N_{max}=2.5, \tau=60, pr=0.01$)	0.0101858	1.18131	-0.246748	0.0555745
“Good AC” ($N_{max}=2.5, \tau=60, pr=0.01$)	0.00988125	1.08033	-0.105267	0.0151403
“Poor AC” ($N_{max}=3, \tau=60, pr=0.03$)	0.0300924	1.20211	-0.311465	0.0798283
Curves without Off-Cycle Power				
“Typical AC” ($N_{max}=2.5, \tau=60$)	0.000352822	1.19199	-0.246716	0.0546566
“Good AC” ($N_{max}=2.5, \tau=60$)	4.28122e-005	1.09001	-0.103863	0.0138504
“Poor AC” ($N_{max}=3, \tau=60$)	0.000582243	1.23565	-0.313841	0.0780726

To implement these curves in DOE-2, use the commands:

```
newPLR CURVE-FIT TYPE=CUBIC COEF=(a,b,c,d) ..
COOL-EIR-FPLR=newPLR ..
```

The “Typical AC” is assumed to have a time constant of 60 seconds at startup, which is typical of values reported in the literature and summarized by Henderson (1992), with values ranging from 30 to 80 seconds. The “Good AC”, which might be representative of a system with a liquid line solenoid or other means of off-cycle refrigerant control, is assumed to have a shorter time constant of 30 seconds.

The maximum thermostat cycling rate is assumed to be 2.5 cycles per hour for the “Typical AC” and “Good AC”. This value was the average measured at 30 Florida homes

by Henderson et al (1991). By comparison Miller and Jaster (1985) measured values of 1.5 to 3 cycles per hour and recommended 3 as the “worst case”. Parken et al (1985) measured values of 1.6, 2.0, and 2.3 cycles/hr in the cooling mode at their 3 test homes. The “Poor AC” is assumed to have a cycling rate of 3 cycle per hour.

The off-cycle power use is expected to be 1% (0.01), or about 40 Watts with a 3 ton unit for the “Typical AC”. This is close to the value of 1.5% assumed by Bonne et al (1980). The “Poor AC” is assumed to be 3% (0.03), or 120 Watts for a 3 ton unit.

If the curves with off-cycle power are used with DOE-2, then the crankcase heater keyword CRANKCASE-HEAT in RESYS should be set to zero (the default value is 50 Watts). Otherwise, the part load curves without off-cycle power can be used and the crankcase heater power can be specified independently. This approach has the advantage of allowing the crankcase heater to only operate whenever the compressor is off and ambient conditions fall below a specified set point (CRANKCASE-MAX-T = 70°F by default).

There is not expected to be much difference between new and existing systems. Most research into part load issues was conducted in the 1970’s and 1980’s, though work at FSEC (Henderson 1990) and other places appears to confirm these earlier findings. While the steady state performance of residential AC and HP systems has improved substantially over the last 10 to 15 years, there is little evidence that part load issues have changed for cycling equipment. The transient response at startup is still expected to be similar, with the exception of systems with liquid line solenoid valves or totally closeable electronic expansion valves, which are expected to respond faster since refrigerant is trapped in the condenser. Thermostat manufacturers also still design for maximum cycling rates of 2 to 3 cycles per hour.

Impact of Part Load Degradation

Table 3 summarizes the impact that the improved part load function has on a typical 1,500 ft² house in Miami. The default RESYS curves add 24% to the annual energy use of the air conditioner. In contrast, the “Poor” and “Good” AC curves (without off cycle power considered) predict that part load efficiency losses are only 4 and 11%, respectively. Annual losses of 5-10% have generally been shown in the literature from field studies and detailed, small time step simulations (Henderson 1992).

Table 3. Impact of Part Load Models on Annual Cooling Energy Use in DOE-2

	Part Load Losses (kWh/yr)	Annual Losses (%)
Default RESYS Curves (SDL-C17)	936	24%
"Poor" AC (w/o off-cycle power, Figure 4)	417	11%
"Good" AC (w/o off-cycle power, Figure 4)	170	4%

Miami House with cooling energy use of 3,831 kWh/yr with part load degradation turned off.

A User Function for Improved Humidity Predictions

DX Coil Model

To accurately predict space humidity levels, the air conditioner model must properly predict the split of sensible and latent capacity over a range of operating conditions. It is generally accepted that the total (sensible and latent) capacity of a DX air conditioner is a function of ambient conditions and the entering wet bulb temperature. However, the sensible and latent portions of the total capacity are a function of the psychrometric state point of the entering air, not just the wet bulb alone.

This aspect of cooling coil performance is best described by the apparatus dew point/bypass factor (ADP/BF) approach. This approach is an analog to the NTU-effectiveness calculations used for sensible-only heat exchanger calculations extended to a cooling and dehumidifying coil. BF, by definition, is one minus the heat exchanger effectiveness for both the latent and sensible calculations. For an air-to-refrigerant heat exchanger (where $C_{\min}/C_{\max}=0$) BF is defined as:

$$BF = 1 - \varepsilon = e^{-NTU} \quad (6)$$

$$NTU = \frac{hA}{\dot{m}\rho c_p} = \text{constant /cfm} \quad (7)$$

Then the leaving temperature and humidity conditions from the coil are found with the ADP and BF as shown below in equations 8 and 9 below. The humidity ratio w_{ADP} corresponds to saturation conditions at the apparatus dew point.

$$T_{leaving} = T_{ADP} \cdot (1 - BF) + BF \cdot T_{entering} \quad (8)$$

$$w_{leaving} = w_{ADP} \cdot (1 - BF) + BF \cdot w_{entering} \quad (9)$$

The DOE-2 documentation (Version 2.1C Reference Manual) describes the cooling algorithms for the DX systems, including RESYS. The DX cooling models use the ADP/BF approach, but the algorithms make the error of assuming that the bypass factor (BF) is a function of the psychrometric conditions entering the coil (i.e., the DB and the WB) in addition to the air flow rate. The variation of BF is specified by the polynomial functions COIL-BF-FT and COIL-BF-FCFM (the default curves are SDL-C31 & SDL-C41). The dependence of BF on DB and WB does not have a physical basis, as shown by equations 6 and 7. However, eliminating the dependence of BF on DB and WB, and making it a fixed value, actually makes the latent/sensible characteristics of RESYS worse. It appears that the original developers have fit this “non-physical” form of the COIL-BF-FT equation to actual data. The variation of BF with DB & WB actually mimics the variation you might expect for the ADP (or TSURF) at least near design conditions. However, less than rational performance is observed at off-design conditions.

To provide a more physically-based model for the DOE-2 RESYS routine, the ACDX.DXDOE subroutine from ASHRAE’s Secondary Toolkit (Brandemuehl et al 1993) was adapted for use as a user-defined function in DOE-2. The model is also described by Henderson, Rengarajan, and Shirey (1992).

The DX AC routine is a “physically rational” model that predicts the latent and sensible capacity of a cooling coil at off-design conditions. It uses the total capacity and efficiency correlations from DOE-2 (PSZ curves from version 2.1c³) along with the bypass factor/apparatus dewpoint (BF/ADP) relations to model performance. The routine uses the nominal capacity, Sensible Heat Ratio (SHR), Energy Efficiency Ratio (EER), at design conditions (i.e., 80°F/67°F entering air, 95°F ambient, 450 cfm/ton) to construct a performance map for off-design conditions.

The original FORTRAN routine included several iteration loops. However, the implementation as a user-defined function was simplified so that iterations were not required. For instance, the user function can not predict proper power use at dry coil conditions (however, this condition should rarely occur in a residential AC system). Also, a simple correlation is used to relate nominal SHR to the nominal BF instead of iterating to find an exact solution. A curve fit that relates enthalpy and temperature at saturation conditions between 25°F and 60°F was also developed to eliminate the need for iterations to find ADP from the leaving enthalpy. These approximations introduce very little error for an air conditioner treating mostly return air in a residential application. The user function is listed in Appendix A.

Moisture Capacitance

In addition to calculating available sensible and latent capacity for the AC cooling coil, the user-defined function also calculates the resulting impact on zone humidity. DOE-2 does not consider moisture storage effects in the building zone. Instead, it calculates the steady-state moisture balance for each hour and solves for the space humidity that achieves a balance. This calculation approach results in humidity levels that are much different than is observed in real buildings and can result in sudden, unrealistic jumps in predicted space humidity conditions. To improve the physical accuracy of simulations, a moisture capacitance model was added to the user-defined function. This model assumes the building has a “lumped” moisture holding capacity, or capacitance (C_a), that is a multiple of the interior air mass. Comparisons to more detailed moisture transport/storage models such as FSEC (Kerestecioglu et al 1989) have shown the simple lumped approach yields results similar to more detailed approaches with a lumped moisture capacitance that is equivalent to 20 times the building air mass. To implement moisture capacitance we first start with the mass balance equation:

$$\frac{dw}{dt} C_a = [\text{Moisture}_{-} \text{gains}] - [\text{AC}_{-} \text{Latent}_{-} \text{Capacity}] \quad (10)$$

Discretizing the differential equation to an hour and using “forward differencing,” the equation for a residence becomes:

$$w'_i = w_i + \frac{[(Q_{lat,space} - PLRC \cdot Q_{AC,L}) + 4770 \cdot cfm_{inf} (w_o - w_i)]}{C_a} \quad (11)$$

³ The version C curves from PSZ were used because they were linear. Other curves for total capacity and efficiency could also be used instead.

Where w_i' is the humidity level for the next hour. Assuming C_a is large, then the humidity changes at a rate less than 1 gr/lb per hour. This makes the errors associated with forward differencing small (i.e., latent capacity and infiltration loads are evaluated with w_i from the current hour to find the space humidity for the next hour). The implementation of the algorithm in the user-defined function is shown in Appendix A.

Impact on Performance

This simple moisture balance approach yields space humidity trends that are consistent with more detailed moisture capacitance models (Shirey and Rengarajan 1996). Figure 5 compares the resulting humidity level with this new modeling approach to the humidity levels predicted with RESYS. The large fluctuations with RESYS are eliminated with the new approach and the humidity levels more closely match what might be expected for a house in Miami. The addition of capacitance makes the daily humidity swings much more inline field-measured performance.

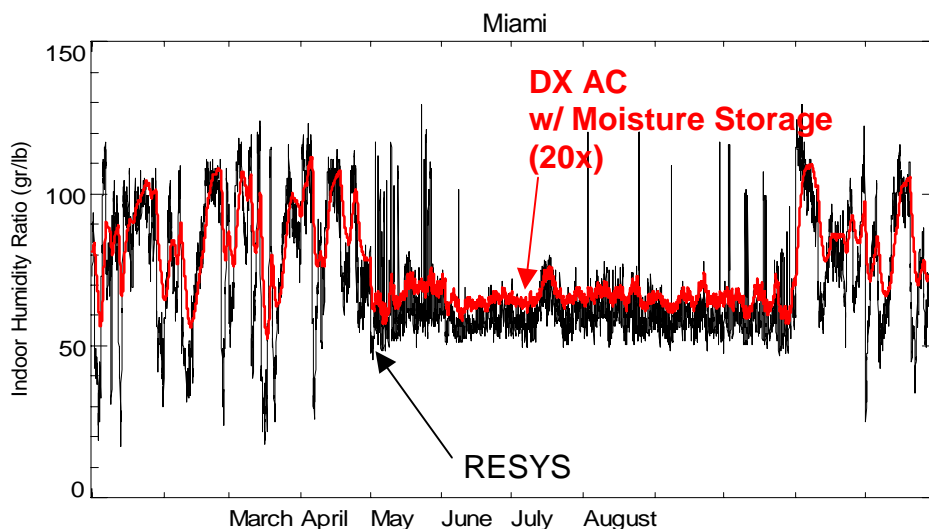


Figure 5 Comparing Zone Humidity Levels with RESYS and DX AC/Moisture Storage Approach

Summary

Two approaches to improve the part load performance of cycling air conditioning systems have been developed. Coefficients are given for improved efficiency degradation curves in DOE-2. The new curves reduce the annual efficiency degradation penalty to the 5-10% range compared to unrealistically large 24% penalty that results with the default RESYS curves. A user defined function for RESYS is also developed with improved cooling coil models and moisture capacitance to provide more accurate predictions of space humidity.

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APPENDIX A – User Defined Function Developed for DOE-2 RESYS

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SUBR-FUNCTIONS
RESYS-5=*DOE_AC_SYS* ..

$***** DOE_AC_SYS *****
FUNCTION NAME = DOE_AC_SYS ..
ASSIGN          $ GLOBAL VARIABLES $
    PLRC=PLRC   pow = SKWQC           $ cooling PLR and AC comp power $
    TM=TM       wml=WM                $ entering coil conditions $
    PATM=PATM   DBT=DBT HUMRAT=HUMRAT $ Ambient Conditions $
    IHR=IHR     IDAY=IDAY IMO=IMO      $ hour, day, month $
    Q=Q         QLSUM=QLSUM CINF=CINF .. $ Sens & Lat Load (Btu/h), infiltration (cfm) $

ASSIGN          $ PARAMETERS $
    qt_rate = 360000   cfm=1200       $ AC Rated Capacity (Btu/h) and supply cfm $
    FLRAREA=1500.     WLHT=8.         $ Used to find air mass $
    SHR_rate=0.75     eir_rate=3413   cd=0.2   $ Rated SHR, Rated EIR, part load cd $
    wm_last=XXX29 ..   $ Undocumented trick to save wm $

CALCULATE ..
C Check if there is a cooling load

    IF (Q .LE. 0.) GOTO 50

C Correlation between rated SHR and rated BF

    BF_rate = 1.9253 - 2.258*SHR_rate

C Find BF at the current flow rate

    cfm_rate = 450.*qt_rate/12000.
    a0 = -ALOG(BF_rate)*cfm_rate
    fcfm = cfm/cfm_rate
    bf = exp(-a0/cfm)

C Entering Conditions

    wbm = WBFS(TM,wm_last,PATM)
    hm = H(TM,wm_last,PATM)

C Use DOE2 functions (2.1c for PSZ) to find total capacity

    qct = qt_rate * (0.418934 + 0.017421*wbm - 0.00617*DBT)*
    & (0.69717199 + 0.39555*fcfm - 0.092727*fcfm**2)

    del_h = qct/(.075*cfm*60.0)

C Use curve fit of TSAT =F(HSAT) [from 25F to 60F] to find tadp

    hadp = hm - del_h/(1.0-bf)
    tadp = -2.13741 + 3.46076*hadp - 0.0501927*hadp**2
    & + 0.000312163*hadp**3

    shr = AMIN( 0.241*(1.0-bf)*(TM-tadp)/del_h , 1.0)

    qcs = qct*AMAX(shr,0.0)
    qcl = qct - qcs

c Find new PLRC

50    PLRC = AMIN( AMAX(Q/(qcs+1.0e-10),0.0),1.0 )

c Moisture Balance
c Assume a moisture capacitance equivalent to 20x interior air mass
c Use forward differencing

    c2 = 1061.0*60.0*PATM/(0.754*(TM+460.))
    cm = FLRAREA*WLHT*0.075*20.*1061.
    wml = wm_last
    & + (c2*CINF*(HUMRAT-wm_last) - (qcl*PLRC -QLSUM))/cm
    wml = AMIN(AMAX(wml,.0005),0.035)

c Find AC Power

    eir1 = eir_rate * (0.282094 - 0.005832*wbm + 0.01167*DBT)
    & * (1.13318 - 0.13318*fcfm)
    pow = (eir1*qct/3.413) * PLRC / (1.0-cd*(1.0-PLRC))*0.001

c WRITE (52,10) IMO,IDAY,IHR,DBT,HUMRAT*7000.,TM,Q/1000.,
c & qcs/1000.,qcl/1000.,
c & PLRC,pow,wm_last*7000.,wml*7000.
C10 FORMAT(2X,3F4.0,11F8.3)
    wm_last = wml
END
C*****
END-FUNCTION ..

```