

**NHANES USE FOR STANDARD AVERAGE BODY WEIGHT**

14 DEC 2020

ABSTRACT

For use in aircraft weight and balance, a method to determine standard average weights for body mass of passengers and crew is defined. The method uses data from the US Centers for Disease Control (CDC) National Health and Nutrition Examination Survey (NHANES) in a manner consistent with guidance from the Federal Aviation Administration (FAA), especially Advisory Circular (AC) 120-27 "Aircraft Weight and Balance Control". Justification for aspects of the definition is discussed. Example values using data from 2003 through 2018 are shown. Results are applicable to Operation Specifications (OpSpecs) approval using survey-derived average weight values.

AUTHORS

William Yingling, Senior Engineer Operations, JetBlue  
Charles Ostick, Senior Flight Operations Engineer, Alaska Airlines

REVIEW COMMITTEE

Kristine Henning, Senior Principle Engineer, Performance Engineering, Delta Air Lines (Chair)  
Mirjeta Beqiri, Ph.D., Professor of Operations Management, Gonzaga University  
Michael Byham, Director Operations Engineering, American Airlines  
Ryan Herzog, Ph.D., Associate Professor of Economics, Gonzaga University  
Rick Martin, Acting Sr. Mgr. Aircraft Performance Engineering, Sun Country Airlines  
Cary Robins, President, American Aeronautics  
Donald Tiemann, Performance Engineer, Engineering, Allegiant Airlines  
Gail Zittel, Senior Engineer - Weight and Balance, The Boeing Company

OWNERSHIP & DISCLAIMER

*This document remains the sole property of the Society of Aircraft Performance and Operations Engineers (SAPOE).*

*SAPOE is a member-based organization promoting the safety and efficiency of flight through knowledge of aircraft performance and weight and balance principles.*

*Except for brief quotations with appropriate citation, copies may be made and distributed only of the complete document including cover page and this disclaimer.*

*SAPOE, the authors, and reviewers assume no liability whatsoever and make no warranty of any kind. Anyone who uses this report, or the source code and data contained in it, is solely responsible for their own operations and for any outcome of such use.*

### PROBLEM STATEMENT

US aviation regulations and Operations Specifications (OpSpecs A097, A098, and A099) require an approved weight and balance control system based on average, assumed, or estimated weight. For passengers and crew, establishing an average body weight by sample weighing is disruptive to operations and is viewed unfavorably as an intrusion on privacy by the US traveling public.

Since 1968, the Federal Aviation Administration (FAA) has established standard average weights for airline passengers and crew and published them in Advisory Circular (AC) 120-27 "Aircraft Weight and Balance Control". Since revision D of AC 120-27, published in 2004, these standard average weights have been based on data published by US Centers for Disease Control (CDC) National Health and Nutrition Examination Survey (NHANES).

Revision F of AC 120-27, published in May 2019, deletes the standard average weight values. FAA Policy Notice 8900.551, published June 11, 2020, makes it mandatory for US operators to establish new weight values consistent with AC 120-27F unless actual weights are used.

AC 120-27F still allows operators to use NHANES to establish new average weights for passengers and crew. However, summary publications by NHANES, including those referred to in FAA guidance, conflict with FAA definitions of terms such as "Child" and "Adult". Furthermore, CDC summary statistics exclude persons, such as pregnant women, who must be considered in average passenger weights determined using AC 120-27F. Thus, these summaries cannot be used directly. NHANES raw data, which does not make these exclusions, must be used. AC 120-27F does not define a specific acceptable method to apply the NHANES raw data.

Therefore, a method to establish average weights for passengers and crew using CDC NHANES raw data, consistent with FAA requirements and definitions, is needed.

### SYSTEMS CONSIDERATIONS

The ability to apply new values in existing weight and balance control systems is also a critical requirement. A general solution for novel system designs is beyond the scope of this paper, but certain aspects of existing systems are well known and should be considered.

Age and gender categories are consistently defined in all revisions of AC 120-27, and these categories are a fundamental assumption in existing weight and balance control systems. These systems often access only the number of persons in each category and may be blocked from accessing more specific personal data even if it is available to an operator's other systems.

Weight and balance control systems differ in their ability to assign average weights to each age and gender category. For example, because past guidance did not assign a weight to infants, some systems can only account for this weight by adjusting the average value used for adults, while others can assign non-zero weight to the number of infants on board.

Changing the design of these systems can introduce significant operational and safety risk and may not be possible to accomplish in the time frame required for initial implementation of new weight values.

Therefore, the needed method must be compatible with existing weight and balance control systems and consistent with past design assumptions.

### METHOD DEFINITION

A method to establish average body weights, meeting the requirements and considerations given, is defined here. No claim is made that this is the only valid method, especially if different requirements and considerations apply. Even within this method, different systems considerations may yield different resulting values.

The method is defined in three parts. The first part defines mathematical steps to determine average weights for a set of FAA-defined passenger categories. The second part defines adjustments that may be required to apply the average weights in a given operator's control systems. The third part adapts the method from Parts 1 and 2 to determine average weights for crewmembers.

How to efficiently accomplish the steps and calculations defined follows in DISCUSSION, along with justification of certain aspects of the method and comparison to other statistical techniques.

#### Part 1, Survey-derived Average Passenger Body Weights from NHANES Data

As allowed by AC 120-27F, values are determined from NHANES raw data. The minimum essential data is:

<b>Table 1: Minimum Required NHANES Data Fields</b>		
Field Name	Data File	Field Description
BMXWT	BMX	Measured body mass in kilograms
BMIWT	BMX	Body mass comment: 1 = Could not obtain; 2 = Exceeds scale capacity; 3 = Respondent Clothed; 4 = Medical appliance included
RIAGENDR	DEMO	Gender: 1 = Male; 2 = Female
RIDAGEYR	DEMO	Age in years at time of examination.
WTMEC2YR	DEMO	Full sample 2-year "weighting factor" to be applied to measured data to relate it to the larger US noninstitutionalized civilian population.
SEQN*	DEMO & BMX	Respondent sequence number: Functions as a unique record key to tie together data from DEMO and BMX files.

*\*SEQN is required only to join records in the BMX file to associated records in the DEMO file.*

As recommended in NHANES Analytic Guidelines and Tutorial, consecutive 2-year cycles are combined into a 4-year cycle. (The most recent 4-year cycle combines 2015-2016 and 2017-2018.) In the formulas below, WTMEC is the 4-year weighting value, which for 2003 – 2018 is half of the WTMEC2YR value in the raw data.

Consistent with AC 120-27E, Appendix 2, Paragraph 1. c., sample data without an examination record, with a missing body mass (BMXWT) value, or coded as clothed when weighed (BMIWT=3) are excluded.

From RIDAGEYR, age categories are defined to comply with FAA definitions in AC 120-27F, and earlier revisions. The "Infant" category is defined as persons who have not yet reached their 2<sup>nd</sup> birthday. The "Child" category is defined as persons with an age greater than or equal to 2 years who have not yet reached their 13<sup>th</sup> birthday. The "Adult" category is defined as persons with an age greater than or equal to 13 years.

For FAA definitions, gender is only considered for "Adults", and is given by RIAGENDR.

For each age and gender category, i.e. “Infant”, “Child”, “Adult Female” and “Adult Male”, the following steps are calculated independently. In formulas (1) and (2),  $n_R$  is the number of non-excluded records in the data set for each category.

Step 1) A raw weighted mean ( $\bar{w}^*$ ) is calculated using the formula:

$$\bar{w}^* = \frac{\sum_{i=1}^{n_R} BMXWT_i \times WTMEC_i}{\sum_{i=1}^{n_R} WTMEC_i} \quad (1)$$

Step 2) Using the raw mean, a raw weighted sample standard deviation ( $\sigma^*$ ) is calculated using the formula:

$$\sigma^* = \sqrt{\frac{\sum_{i=1}^{n_R} WTMEC_i \times (BMXWT_i - \bar{w}^*)^2}{\frac{(n_R - 1)}{n_R} \times \sum_{i=1}^{n_R} WTMEC_i}} \quad (2)$$

Step 3) Consistent with AC 120-27E, Appendix 2, sample data with BMXWT more than twice  $\sigma^*$  either greater than or less than  $\bar{w}^*$  are excluded to form a filtered data set. (See DISCUSSION for elaboration.) In formulas (3) and (4),  $n_f$  is the number records in the filtered data set for each category.

Step 4) From the filtered data set, a weighted mean ( $W$ ) is calculated using the formula:

$$W = \frac{\sum_{i=1}^{n_f} BMXWT_i \times WTMEC_i}{\sum_{i=1}^{n_f} WTMEC_i} \quad (3)$$

Step 5) From the filtered data set, a weighted sample standard deviation ( $\sigma$ ) is calculated using the formula:

$$\sigma = \sqrt{\frac{\sum_{i=1}^{n_f} WTMEC_i \times (BMXWT_i - W)^2}{\frac{(n_f - 1)}{n_f} \times \sum_{i=1}^{n_f} WTMEC_i}} \quad (4)$$

Step 6) Due to NHANES sample design, the tolerable error percentage ( $e$ ) defined in AC 120-27F should not be used to determine confidence intervals for NHANES averages. However, if an operator is required to demonstrate these values, they may be calculated using AC 120-27F, paragraph 3.3.3:

$$e = \frac{1.96 \times \sigma \times 100}{W \times \sqrt{n_f}} \quad (5)$$

The survey-derived average passenger body weights are the  $W$  values defined for Infant, Child, Adult Female, and Adult Male, labeled  $W_I$ ,  $W_C$ ,  $W_F$ , and  $W_M$  respectively.

## Part 2, Weight Adjustments for System-specific Implementation

For application to existing systems, further adjustments may be required. In the following definitions,  $[W]$  is used to represent the adjusted value used in operational systems in place of the  $W$  value calculated as defined in Part 1.

Determining these adjustments requires definition of the following quantities:

- $N_I$  is the total number of Infants boarded
- $N_C$  is the total number of Children boarded
- $N_A$  is the total number of Adults boarded
- $N_F$  is the total number of Adults boarded, known to be Female
- $N_M$  is the total number of Adults boarded, known to be Male

$N_X$  is the total number of Adults boarded, not known to be Female or Male

$$N_A = N_F + N_M + N_X \quad (6)$$

Values for these quantities should be determined for the operation to which the adjusted  $[W]$  will be applied. For example, they might be measured by counting on representative flights for a representative time period, or estimated from existing data. Estimates for  $N_F$ ,  $N_M$ ,  $N_G$ , and  $N_I$  may also be obtained from the NHANES filtered data set defined in Part 1, Step 3, by dividing the total WTMEC for each category by the total WTMEC for all categories, using the formula:

$$N = \frac{\sum_{i=1}^{n_f} WTMEC_i}{\sum_{F,M,C,I} (\sum_{i=1}^{n_f} WTMEC_i)} \quad (7)$$

Additionally:

M:F is the assumed Male to Female ratio for Adults when gender is not identified; and  
 $[W_X]$  is the operationally assumed weight for an Adult when gender is not identified.

Values for M:F are sometimes assigned to an operator by the FAA. For example, "50:50" is stated in the FAA template for Operation Specifications A099, but "60:40" or other ratios may be given in other contexts. (When M:F = 60:40, M = 60 and F = 40.) If M:F is not defined by the FAA, use:

$$M:F = N_M : N_F \quad (8)$$

Adjusted values  $[W]$  must ensure that total weight calculated by the operational system is equal to total weight calculated assuming M:F and the  $W$  values from Part 1, which is stated in the following identity: (9)

$$N_X [W_X] + N_M [W_M] + N_F [W_F] + N_C [W_C] + N_I [W_I] = \left(N_M + \frac{M}{M+F} N_X\right) W_M + \left(N_F + \frac{F}{M+F} N_X\right) W_F + N_C W_C + N_I W_I$$

Any set of  $[W]$  values that preserve the required identity (9) is consistent with this method definition.

For a common set of system constraints, where gender is always unknown ( $N_M = N_F = 0$  implying  $N_X = N_A$ ) and infants cannot be assigned a weight (implying  $[W_I] = 0$ ), the required identity is satisfied without adjusting Child weights (implying  $[W_C] = W_C$ ) if the Adult weight is adjusted according to:

$$[W_X] = \frac{M W_M + F W_F}{M+F} + \frac{N_I}{N_A} W_I \quad (10)$$

Furthermore, if M:F is required to be 50:50 with  $[W_I] = 0$  and  $[W_C] = W_C$  then:

$$[W_X] = \frac{1}{2}(W_M + W_F) + \frac{N_I}{N_A} W_I \quad (11)$$

Note that if the operational system can assign a weight to infants and M:F = 50:50, then the required identity (9) could also be satisfied with  $[W_I] = W_I$  and  $[W_C] = W_C$  by

$$[W_X] = \frac{1}{2}(W_M + W_F). \quad (12)$$

### Part 3, Survey-derived Average Crewmember Body Weights from NHANES Data

Definitions in Part 1 and Part 2 are applied to NHANES raw data, except with different age and gender categories.

Since the FAA defined age categories of “Infant”, “Child”, and “Adult” do not apply to crewmembers, a “Pilot” category is defined as persons at least 23, and less than 65, years of age and a “Flight Attendant” category is defined as persons at least 21 years of age.

The survey-derived average crewmember body weights are the  $W$  values defined for Pilot Female, Pilot Male, Flight Attendant Female and Flight Attendant Male, labeled  $W_{P,F}$ ,  $W_{P,M}$ ,  $W_{FA,F}$ , and  $W_{FA,M}$  respectively.

If the gender of crewmembers actually on each flight is not known, then  $[W_P]$  for Pilots and  $[W_{FA}]$  for Flight Attendants must be determined. This requires a Male to Female ratio for Pilots ( $M_P:F_P$ ) and a Male to Female ratio for Flight Attendants ( $M_{FA}:F_{FA}$ ). Formula (10) is restated for Pilots and Flight Attendants as

$$[W_P] = \frac{M_P W_{P,M} + F_P W_{P,F}}{M_P + F_P} \quad \text{and} \quad [W_{FA}] = \frac{M_{FA} W_{FA,M} + F_{FA} W_{FA,F}}{M_{FA} + F_{FA}} \quad (13)$$

For crew members other than pilots and flight attendants, the weight for the category that more closely aligns with crew members’ duties should be used. For example, an operator might use “Pilot” weights for Flight Engineers and “Flight Attendant” weights for supernumeraries or non-certificated crewmembers.

## DISCUSSION

The calculations required by METHOD DEFINITION can be accomplished in any available calculation tool, including a spreadsheet. However, more thorough analysis requires tools meant specifically for statistical work. A text-based tool allows inclusion of the analysis steps and results in this paper. “R”, an analysis tool accessible to many data analysts and operations engineers, is therefore used here.

It is hoped the style used will be readable by those accustomed to working with data, even without extensive background in the language. Full R scripts, with instructions for getting started in R, are included in an Appendix. Code fragments in this section should be executed in the order presented. For clarity, fixed-width font is used with executable code shown in blue, descriptive comments in green, and any output generated in orange, starting with:

```
# The 'dplyr' library simplifies and improves readability, especially with
# 'mutate()' to add calculated columns to a data table,
# 'group_by()' to define categories within the data, and
# 'summarise()' [note spelling] to calculate summary statistics
library(dplyr)
```

Note that the symbol “%>%” means: to the result of the commands so far on the left, next apply the action to the right.

NHANES is a rigorous scientific study of health and nutrition among US, non-institutionalized, civilian population whose scope includes numerous body size metrics. Documentation for each NHANES data release should be reviewed whenever the data is used.

- NHANES is described for the public at [[www.cdc.gov/nchs/nhanes/about\\_nhanes.htm](http://www.cdc.gov/nchs/nhanes/about_nhanes.htm)].
- Academically rigorous documentation, including Sample Design, Weighting Process, and Analytical Guidelines, is available in a series of papers at [[www.cdc.gov/nchs/nhanes/analyticguidelines.aspx](http://www.cdc.gov/nchs/nhanes/analyticguidelines.aspx)].
- A technical Tutorial for data analysis which discusses background concepts and includes practical examples is at [[www.cdc.gov/nchs/nhanes/tutorials](http://www.cdc.gov/nchs/nhanes/tutorials)].

- The NHANES Variable Keyword Search at [[www.cdc.gov/nchs/nhanes/search](http://www.cdc.gov/nchs/nhanes/search)] can be used to determine how fields of interest are identified in the NHANES data.

*Note: NHANES generally (not exclusively) uses “weight” to mean a numeric scaling factor, as in the common meaning of “weighted average”, rather than a measure of gravitational mass. To avoid confusion, this section carefully uses “mass” as the object of gravity and follows the NHANES convention for “weight” despite this paper elsewhere using US aviation vernacular, including “weight and balance” for the effect of gravity.*

Using documentation above, details referred to in [METHOD DEFINITION](#) and [DISCUSSION](#) were confirmed to be consistent from 2003 through 2018. Data prior to 2003 uses inconsistent definitions and this method cannot be used without changes.

NHANES uses “MEC” to refer to CDC Mobile Examination Centers in which body measurements are taken. Body mass measurements BMXWT include undergarments and an exam gown, but not other clothing, unless an exception is coded using BMIWT.

In addition to the minimal data listed in Table 1, this analysis will use values identified in Table 2.

Field Name	Data File	Field Description
SDDSRVYR	DEMO	This variable represents the two-year data release cycle number.
RIDEXPRG	DEMO	Pregnancy status for females between 20 and 44 years of age at the time of MEC exam.
SDMVPSU	DEMO	Masked variance unit pseudo-primary sampling unit variable; used for variation and error analysis
SDMVSTRA	DEMO	Masked variance unit pseudo-stratum variable; used for variation and error analysis

When working with multiple cycles, SDDSRVYR identifies the cycle in each record. File names use the corresponding letter of the alphabet. For example, raw data for the 2017-2018 is identified in the data by SDDSRVYR = 10 and the letter “J”, the tenth letter of the alphabet.

The required data files, DEMO and BMX, are freely available for download from the NHANES website. For example:

DEMO: [www.cdc.gov/nchs/nhanes/2017-2018/DEMO\\_J.XPT](http://www.cdc.gov/nchs/nhanes/2017-2018/DEMO_J.XPT)

BMX: [www.cdc.gov/nchs/nhanes/2017-2018/BMX\\_J.XPT](http://www.cdc.gov/nchs/nhanes/2017-2018/BMX_J.XPT)

The following R commands download the required data files. Once published, these files do not change, so this only needs to be done once and not every time the analysis is repeated.

```
# Define a helper to keep cycle number and year conventions consistent
cycleYears <- function( cycle2or4, cycleNumbers )
{
  yr <- 2000+2*(cycleNumbers-1)
  sprintf('%s-%s',yr-(cycle2or4-1),yr)
}

# Define a helper to download needed data files to the working directory
downloadDataCycles <- function( cycleNumbers )
{
```

Copyright © 2020, Society of Aircraft Performance and Operations Engineers (SAPOE).

See page 1 for rights claimed, conditions and disclaimer.

```

for( cn in cycleNumbers ){
  files <- sapply(c( 'DEMO_%s.XPT', 'BMX_%s.XPT' ), sprintf, LETTERS[cn])
  for( f in files ){
    download.file(
      sprintf('https://wwwn.cdc.gov/nchs/nhanes/%s/%s', cycleYears(2,cn), f),
      f, mode='wb'
    )
  }
}
}

# Change the working directory to one the current user can save files locally
setwd( 'C:/NHANES/' )

# Download XPT files for the cycles of interest
# Do this just once (ever) and then comment out this line
downloadDataCycles(3:10)

```

The format of these files is SAS Transport [.XPT]. Many statistical software packages, including R, can read this format directly, which is preferred. Otherwise, the CDC provides a link to a free universal SAS file viewer [[wwwn.cdc.gov/nchs/nhanes/sasviewer.aspx](http://wwwn.cdc.gov/nchs/nhanes/sasviewer.aspx)] which may be used to paste the data into more basic programs, such as a spreadsheet.

The following R commands read the required data files and load just the variables in Table 1 and Table 2 into memory. The qualification '!is.na(WTMEC2YR)' loads BMX and DEMO data for every person in the sample not missing an examination record.

```

# The 'foreign' library reads SAS Transport(XPT) files
library(foreign)

# Define a helper to load the needed data columns files in the working directory
importDataCycles <- function( cycleNumbers )
{
  allData <- data.frame(NULL)
  for( cn in cycleNumbers ){
    allData <- bind_rows(allData, left_join(
      subset(
        select(
          read.xport(sprintf('DEMO_%s.XPT', LETTERS[cn])),
          # only needed columns from the DEMO file here
          'SEQN', 'SDDSRVYR', 'RIDAGEYR', 'RIAGENDR', 'RIDEXPRG',
          'SDMVSTRA', 'SDMVPSU', 'WTMEC2YR'
        ),
        !is.na(WTMEC2YR)), # exclude participants without an exam record
      select(
        read.xport(sprintf('BMX_%s.XPT', LETTERS[cn])),
        # only needed columns from the BMX file here
        'SEQN', 'BMXWT', 'BMIWT'
      ),
      by = 'SEQN'
    )
  }
}

```



```

    )
  }
  return(allData)
}

# Load the data into memory once each session
RawData <- importDataCycles(3:10)

```

WTMEC2YR is a weight factor determined from the inverse probability that each person in the 2-year cycle exam data sample was selected from the population. It takes into account complex survey design, including oversampling, non-response, and post-stratification adjustments and is based on population data from the US Census Bureau. This weighting factor should be applied to relate data from MEC sample records in a 2-year cycle to the study population.

NHANES Analytic Guidelines recommend merging consecutive 2-year cycles into 4-year cycles to improve accuracy. To relate data from MEC sample records in a 4-year cycle to the study population a different weight factor, called WTMEC4YR, must be calculated. NHANES samples for years 2003-2018 were designed so that the correct values of WTMEC4YR are exactly half of the published value of WTMEC2YR. This is illustrated in [\[https://wwwn.cdc.gov/nchs/nhanes/tutorials/module3.aspx\]](https://wwwn.cdc.gov/nchs/nhanes/tutorials/module3.aspx). While average passenger body mass  $W$  is defined using a 4-year cycle with  $WTMEC = WTMEC4YR = WTMEC2YR/2$ , for data comparison formulas (1) thru (4) and (7) are also used here for 2-year cycles with  $WTMEC = WTMEC2YR$ .

NHANES Analytical Guidelines also describes the sample selection method as “stratified, multi-stage probability sample design” for which exact mathematical formulas for variance estimates do not exist. Instead, the National Center for Health Statistics (NCHS) recommends “Taylor series linearization methods for variance estimation in analyses of NHANES data for 2-year cycles or combined cycles”. In R, this is provided by `library(survey)`. This method of variance estimation requires information regarding the survey strata and sampling units which indicate how the sample persons were selected from the study population. These survey design parameters are encoded in the SDMVSTRA and SDMVPSU fields. When analyzing any subset of the sample, this variance estimation requires design parameters for both the records included and those excluded. Thus, when excluding data from any analysis, new fields are added to indicate whether each record is part of the included subset instead of deleting the excluded records.

In R, the required indicator fields, along with age and gender categories as defined by FAA (and by CDC for comparison) are defined first.

```

# Create a copy with calculated columns added (preserve RawData)
# Don't remove any rows, or survey design parameters are lost.
AllData <- mutate(RawData,
  # 'one' is needed to count records
  one = 1,

  # combine 2-year cycles as recommended by Analytical Guidelines
  cycle4 = 2*ceiling(SDDSRVYR/2),
  WTMEC4YR = WTMEC2YR/2,

  # an alias to ease comparing 4- and 2-year cycles
  cycle2 = SDDSRVYR,
  # use with WTMEC2YR

```

```

# make it easy to work in US units
WTLB = BMXWT / 0.45359237, # kg to lb

# use cut(right=FALSE) to build age intervals that include the lower
# and exclude the upper boundary
#
# for FAA, Infants have not yet reached 2nd birthday.
# Children have not yet reached 13th birthday.
faaAge = cut(RIDAGEYR, breaks=c(-Inf, 2, 13, Inf),
             right=FALSE, labels=c('Infant', 'Child', 'Adult')),
gender=factor(ifelse(RIDAGEYR>12,RIAGENDR,0),labels=c('None', 'Male', 'Female')),
#
# for comparison to CDC/NHSR, Adults are age 20 and over
cdcAge = cut(RIDAGEYR, breaks=c(-Inf, 20, Inf),
             right=FALSE, labels=c('Child', 'Adult'))
)

# Add columns used to filter the data, like 'inAnalysis' in tutorials
# Don't remove any rows, or survey design parameters are lost.
AllData <- mutate(AllData,
  NotMissing = ( !is.na(BMXWT) ),
  NotClothed = ( !is.na(BMXWT) & (is.na(BMIWT) | 3!=BMIWT) ),
  NotPregnant = ( !is.na(BMXWT) & (is.na(RIDEXPRG) | 1!=RIDEXPRG) )
  # for RIDEXPRG, 1=Yes, 2=No, 3=Unknown, Missing => No
)

```

Then, the NHANES survey design parameters are defined for 2-year and 4-year cycles. `svydesign` and `svyby` are library functions which apply NCHS recommended variance estimates using the survey design parameters. User defined function `svyWTLB` simply saves retyping commands to use these functions, especially to combine weighted statistics with unweighted sample size. `fmtWTLB` provides a consistent format to simplify comparing results.

```

library(survey)

# Define survey design parameters for overall dataset
NHANES2 <- svydesign(data=AllData,
  id=~SDMVPSU, strata=~SDMVSTRA, weights=~WTMEC2YR, nest=TRUE)
NHANES4 <- svydesign(data=AllData,
  id=~SDMVPSU, strata=~SDMVSTRA, weights=~WTMEC4YR, nest=TRUE)

# Define a function to get the weighted mean and standard error estimate
# using survey design parameters and join to the unweighted count
svyWTLB <- function(design, byFormula)
{
  # Get mean, stderr, and unweighted sample size
  c <- svyby( ~WTLB, byFormula, design, unwtd.count )
  p <- svyby( ~WTLB, byFormula, design, svymean )
  # select(c,-se) excludes the standard error of the count
  # suppressMessages() just hides output of the implicit "by" columns
  suppressMessages( outData <- left_join(select(c,-se), p) )
  return(outData)
}

```

```

}

# Define a function to get consistent output formatting
fmtWTLB <- function(cycleData)
{
  c2 <- match("cycle2",names(cycleData))
  if(!is.na(c2)) cycleData <- mutate(cycleData,
                                     Years=cycleYears(2,cycle2), .after=cycle2)
  c4 <- match("cycle4",names(cycleData))
  if(!is.na(c4)) cycleData <- mutate(cycleData,
                                     Years=cycleYears(4,cycle4), .after=cycle4)
  print.data.frame( cycleData %>% mutate_if(is.numeric,round,digits=1) )
}

```

The R commands so far can be checked by comparing their results values published by the CDC in National Health Statistics Reports (NHSR). Report Number 122 gives summary data from NHANES and details how these were determined using 2-year cycles, CDC age categories, and excluding pregnant women. Table 2 is reproduced here, and the highlighted results recalculated with the following R command.

**Table 2. Mean weight (pounds) among men and women aged 20 and over, by survey years, age group, and race and Hispanic origin: United States, 1999–2016**

Sex, age, and race and Hispanic origin	Survey years								
	1999–2000	2001–2002	2003–2004	2005–2006	2007–2008	2009–2010	2011–2012	2013–2014	2015–2016
<b>Men</b>									
20 and over <sup>1</sup> (crude)	189.1 (1.5)	191.8 (1.1)	193.5 (1.1)	196.0 (1.7)	194.7 (1.4)	196.3 (1.4)	194.4 (1.4)	197.0 (1.2)	197.8 (1.9)
20 and over <sup>2</sup> (age adjusted)	189.4 (1.5)	191.2 (1.2)	193.3 (1.1)	195.7 (1.7)	194.6 (1.4)	196.2 (1.3)	194.3 (1.4)	197.0 (1.3)	197.9 (2.0)
<i>Mean (standard error)</i>									
Age group:									
20–39	185.8 (2.0)	188.1 (1.6)	190.4 (2.3)	192.1 (2.3)	189.9 (2.1)	193.1 (1.9)	190.7 (1.9)	194.4 (2.7)	196.9 (3.1)
40–59	194.3 (2.6)	197.7 (1.5)	198.4 (2.0)	203.0 (2.5)	199.9 (2.0)	202.0 (2.4)	200.4 (2.3)	200.7 (1.8)	200.9 (2.0)
60 and over	187.8 (1.7)	186.2 (1.5)	190.0 (1.7)	190.0 (1.4)	194.0 (1.8)	192.1 (1.5)	190.5 (2.5)	195.4 (2.7)	194.7 (1.9)
Race and Hispanic origin <sup>3</sup> :									
Non-Hispanic white	192.3 (1.7)	194.6 (1.3)	196.3 (1.3)	197.8 (1.6)	198.2 (1.3)	199.5 (1.4)	196.7 (1.8)	200.2 (2.0)	202.2 (2.4)
Non-Hispanic black	188.7 (1.7)	190.8 (1.9)	196.4 (3.8)	201.0 (1.9)	197.1 (2.0)	201.1 (2.4)	198.4 (2.1)	199.3 (2.6)	197.7 (2.6)
Non-Hispanic Asian	---	---	---	---	---	---	159.1 (1.6)	161.7 (1.4)	161.1 (1.3)
Hispanic <sup>4</sup>	---	---	---	---	---	---	187.2 (1.8)	189.6 (3.1)	190.5 (1.9)
Mexican-American	177.9 (1.6)	177.5 (1.7)	180.1 (2.1)	179.3 (2.0)	184.0 (2.7)	185.6 (2.8)	189.5 (2.5)	191.9 (2.3)	190.4 (2.5)
<b>Women</b>									
20 and over <sup>1</sup> (crude)	163.6 (1.7)	162.9 (1.2)	164.2 (1.6)	165.1 (1.7)	166.1 (1.3)	166.2 (0.9)	167.1 (1.3)	169.8 (1.3)	170.5 (1.7)
20 and over <sup>2</sup> (age adjusted)	163.8 (1.7)	162.8 (1.3)	164.1 (1.7)	164.8 (1.7)	166.3 (1.4)	166.1 (1.0)	167.2 (1.3)	170.1 (1.4)	170.8 (1.7)
Age group:									
20–39	161.9 (2.1)	158.9 (2.1)	160.6 (2.4)	160.5 (2.4)	166.6 (2.7)	164.3 (2.1)	165.1 (1.8)	169.4 (1.7)	167.6 (1.9)
40–59	169.4 (2.9)	168.6 (2.4)	171.0 (2.5)	172.3 (2.7)	170.2 (1.8)	167.8 (1.5)	172.3 (1.7)	175.2 (2.7)	176.4 (3.0)
60 and over	157.9 (1.3)	160.3 (1.4)	159.0 (1.3)	160.2 (1.9)	159.5 (1.1)	166.6 (1.3)	162.5 (2.3)	163.1 (1.7)	166.5 (2.6)
Race and Hispanic origin <sup>3</sup> :									
Non-Hispanic white	161.9 (2.2)	162.2 (1.3)	162.6 (2.0)	164.1 (1.9)	165.8 (2.2)	165.0 (1.2)	167.1 (1.9)	170.4 (1.8)	170.9 (2.1)
Non-Hispanic black	185.9 (2.3)	179.3 (1.9)	185.1 (2.3)	184.1 (2.3)	185.0 (3.0)	189.5 (1.9)	190.0 (2.2)	190.0 (1.9)	186.4 (2.4)
Non-Hispanic Asian	---	---	---	---	---	---	131.8 (1.7)	131.0 (1.8)	132.4 (1.1)
Hispanic <sup>4</sup>	---	---	---	---	---	---	162.4 (1.3)	166.3 (2.1)	169.0 (1.6)
Mexican-American	157.5 (2.2)	157.5 (2.1)	162.7 (2.7)	160.8 (2.3)	160.8 (0.9)	161.6 (1.4)	165.0 (2.4)	170.0 (2.2)	171.9 (1.5)

--- Data not available.  
<sup>1</sup>Includes other races not shown separately.  
<sup>2</sup>Age adjusted to the projected 2000 U.S. census population using age groups 20–39, 40–59, and 60 and over.  
<sup>3</sup>Includes Mexican-American persons.  
 SOURCE: NCHS, National Health and Nutrition Examination Survey, 1999–2016.

```

# confirm definitions above here are correct by comparing to
# https://www.cdc.gov/nchs/data/nhsr/nhsr122-508.pdf

fmtWTLB( svyWTLB( subset( NHANES2, NotPregnant ), ~cycle2+cdcAge+gender ) ) %>%
  filter(cdcAge=='Adult'&cycle2<10) # NHSR Table 2 has Adults only

```

	cycle2	Years	cdcAge	gender	counts	WTLB	se
1	3	2003-2004	Adult	Male	2247	193.5	1.0
2	4	2005-2006	Adult	Male	2242	196.0	1.7
3	5	2007-2008	Adult	Male	2755	194.7	1.4
4	6	2009-2010	Adult	Male	2896	196.3	1.4
5	7	2011-2012	Adult	Male	2591	194.4	1.4

6	8	2013-2014	Adult	Male	2645	197.0	1.2
7	9	2015-2016	Adult	Male	2584	197.8	1.9
8	3	2003-2004	Adult	Female	2201	164.2	1.6
9	4	2005-2006	Adult	Female	2129	165.1	1.7
10	5	2007-2008	Adult	Female	2805	166.2	1.3
11	6	2009-2010	Adult	Female	3039	166.3	0.9
12	7	2011-2012	Adult	Female	2602	167.1	1.3
13	8	2013-2014	Adult	Female	2823	169.8	1.3
14	9	2015-2016	Adult	Female	2757	170.5	1.7

(In this and subsequent output, the first column is an automatically generated row number for reference only.)

Data in the NHSR report was calculated using SUDAAN software rather than R, yet comparing the output above to highlighted values in the published table shows an almost exact match, with no difference greater than 0.1 lb. This substantiates that the raw data is loaded correctly, and calculated mean and standard error values are consistent with recommended practice by CDC/NHANES and NCHS. The same command, adjusted only to use 4-year cycles, FAA age categories, and not exclude pregnant females, produces the following values:

```
# run the same formula with faaAge, and not excluding pregnant females
#fmtWTLB( svyWTLB( subset( NHANES2, NotPregnant ), ~cycle2+cdcAge+gender ) )
  fmtWTLB( svyWTLB( subset( NHANES4, NotClothed ), ~cycle4+faaAge+gender ) )
```

	cycle4	Years	faaAge	gender	counts	WTLB	se
1	4	2003-2006	Infant	None	1578	21.5	0.2
2	6	2007-2010	Infant	None	1451	21.0	0.2
3	8	2011-2014	Infant	None	1212	20.9	0.2
4	10	2015-2018	Infant	None	1172	21.0	0.2
5	4	2003-2006	Child	None	4352	66.8	0.6
6	6	2007-2010	Child	None	4373	66.3	0.7
7	8	2011-2014	Child	None	4501	67.5	0.8
8	10	2015-2018	Child	None	3890	67.5	0.6
9	4	2003-2006	Adult	Male	6233	189.9	0.9
10	6	2007-2010	Adult	Male	6603	191.0	1.0
11	8	2011-2014	Adult	Male	6152	191.6	0.9
12	10	2015-2018	Adult	Male	5940	194.7	1.3
13	4	2003-2006	Adult	Female	6593	161.7	1.2
14	6	2007-2010	Adult	Female	6741	163.0	0.8
15	8	2011-2014	Adult	Female	6407	165.5	0.9
16	10	2015-2018	Adult	Female	6235	167.9	1.2

This method might not be obviously the same as [METHOD DEFINITION](#), but in fact the definitions are identical for mean values. (Differences in variance estimates are discussed below.) To demonstrate this, a manual implementation of Steps (1) and (2) to calculate  $\bar{w}^*$  (called ws in code) and  $\sigma^*$  (called ss in the code) follow.

```
# A copy of the data excluding clothed sample participants
Step0 <- AllData %>% filter(NotClothed)

# Calculate a raw weighted mean, w-bar-star (ws) using Formula (1)
Step1 <- mutate(Step0, ww=WTLB*WTMEC4YR) %>%
  group_by(faaAge,gender,cycle4) %>%
```

```

summarise( ws=sum(ww)/sum(WTMEC4YR) )

# Sum weighted residuals (wr) for each row to get sigma-star (ss) using Formula (2)
Step2 <- left_join(Step0,Step1,by=c('faaAge','gender','cycle4'))      %>%
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 )                                %>%
  group_by(faaAge,gender,cycle4,ws)                                %>%
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*ss/ws/sqrt(n) )

fmtWTLB(Step2 %>% select(cycle4,faaAge,gender,n,ws,ss,'te%'))

```

	cycle4	Years	faaAge	gender	n	ws	ss	te%
1	4	2003-2006	Infant	None	1578	21.5	5.6	1.3
2	6	2007-2010	Infant	None	1451	21.0	5.5	1.4
3	8	2011-2014	Infant	None	1212	20.9	5.7	1.5
4	10	2015-2018	Infant	None	1172	21.0	5.6	1.5
5	4	2003-2006	Child	None	4352	66.8	33.4	1.5
6	6	2007-2010	Child	None	4373	66.3	33.6	1.5
7	8	2011-2014	Child	None	4501	67.5	34.8	1.5
8	10	2015-2018	Child	None	3890	67.5	33.1	1.5
9	4	2003-2006	Adult	Male	6233	189.9	44.7	0.6
10	6	2007-2010	Adult	Male	6603	191.0	45.8	0.6
11	8	2011-2014	Adult	Male	6152	191.6	45.9	0.6
12	10	2015-2018	Adult	Male	5940	194.7	49.2	0.6
13	4	2003-2006	Adult	Female	6593	161.7	44.1	0.7
14	6	2007-2010	Adult	Female	6741	163.0	43.9	0.6
15	8	2011-2014	Adult	Female	6407	165.5	45.7	0.7
16	10	2015-2018	Adult	Female	6235	167.9	46.5	0.7

The calculated values for the raw weighted mean are identical to those calculated immediately above. The FAA Tolerable Error (te%) calculated from  $\sigma^*$  is expressed as a percentage of the mean, while the Taylor-series based standard error (se) has units of mass, so these values should not be compared directly.

Consistent with the purpose of NHANES as a health and nutrition study, this data represents individuals across the entire spectrum of health and nutrition conditions, including the most unhealthy and malnourished. Individuals at the extremes of such a health spectrum are likely to be underrepresented in, if not completely absent from, the flying public. Thus, when applied to aviation, extreme values in NHANES data should have a reduced effect. Care should be taken that both low and high outliers are excluded.

A standard technique to reduce the effect of extreme values is to exclude sample values which are more than a multiple of the sample standard deviation from the sample mean; twice the standard deviation is common practice.

In certain contexts, it is important to account for possible variation between the survey-derived average value and values for groups of actual passengers. Operational systems which account for this variation have assumed a  $2\sigma^*$  filtering technique because of its publication in AC 120-27, revisions D and E. Choosing a consistent definition avoids having to redesign these systems. An example is AC 120-27F, Appendix D, which relies on the sigma value derived using AC 120-27E, Appendix 2.

The estimates used for sample standard deviation,  $\sigma^*$  and  $\sigma$ , are technically exact only for a simple, independent random sample selection. However, academic reviewers advised that such a simplification can be used for NHANES. This approximation has the advantage that it can be represented exactly by a formula and implemented using any calculation tool without requiring specific software.

A disadvantage which may prove pertinent in the future, when changes in NHANES data will have to be evaluated compared to the current values, is that formulas (2), (4), and (5) do not fully reflect the design of NHANES sample selection. They are thus unreliable for determining whether changes between cycles are due to randomness in sample selection or due to changes in the population. Such an analysis should be accomplished using recommended practices by CDC/NHANES and NCHS.

The defined method of filtering extreme values from the data is implemented in the following commands.

```
# Filter by excluding values more than twice ss from the raw mean
Step3 <- left_join(Step0,Step2,by=c('faaAge','gender','cycle4'))      %>%
  filter( 2 >= abs((WTLB-ws)/ss) )

# Calculate a weighted mean, W using Formula (3)
Step4 <- mutate(Step3, wr=WTLB*WTMEC4YR)                             %>%
  group_by(faaAge,gender,cycle4)                                     %>%
  summarise( W=sum(wr)/sum(WTMEC4YR) )

# Sum weighted residuals (wr) for each row to get sigma (S) using Formula (4)
Step5 <- left_join(Step3,Step4,by=c('faaAge','gender','cycle4'))    %>%
  mutate( wr=WTMEC4YR*(WTLB-W)^2 )                                  %>%
  group_by(faaAge,gender,cycle4,W)                                 %>%
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*S/W/sqrt(n) )

fmtWTLB(Step5 %>% select(cycle4,faaAge,gender,n,W,S,'te%'))
```

	cycle4	Years	faaAge	gender	n	W	S	te%
1	4	2003-2006	Infant	None	1494	21.8	5.1	1.2
2	6	2007-2010	Infant	None	1382	21.1	4.9	1.2
3	8	2011-2014	Infant	None	1167	20.9	5.2	1.4
4	10	2015-2018	Infant	None	1118	21.1	5.0	1.4
5	4	2003-2006	Child	None	4087	62.0	26.4	1.3
6	6	2007-2010	Child	None	4156	61.5	26.7	1.3
7	8	2011-2014	Child	None	4303	62.4	26.9	1.3
8	10	2015-2018	Child	None	3692	62.9	26.3	1.3
9	4	2003-2006	Adult	Male	5940	186.1	36.1	0.5
10	6	2007-2010	Adult	Male	6291	186.5	37.3	0.5
11	8	2011-2014	Adult	Male	5879	186.8	37.1	0.5
12	10	2015-2018	Adult	Male	5689	189.6	40.5	0.6
13	4	2003-2006	Adult	Female	6321	155.9	35.1	0.6
14	6	2007-2010	Adult	Female	6435	157.3	35.3	0.5
15	8	2011-2014	Adult	Female	6125	159.3	36.4	0.6
16	10	2015-2018	Adult	Female	5937	161.6	37.2	0.6

These are the survey-derived averages for body mass calculated in accordance with Part 1 of [METHOD DEFINITION](#).

In Part 2 of METHOD DEFINITION, the operator should select the combinations of adjustments, or none, which are most appropriate to their systems. For some systems these may be calculated for each flight, season, or other period, while some systems require static adjustments updated only when OpSpecs are revised.

Estimated values of  $N$  for each FAA age/gender category can be determined from NHANES raw data.

```
# Calculate population fractions using Formula (7)
NN <- left_join(
  Step3 %>% group_by(cycle4,faaAge,gender) %>% summarise( Nc=sum(WTMEC4YR) ),
  Step3 %>% group_by(cycle4,
                    ) %>% summarise( Nt=sum(WTMEC4YR) ),
  by=c('cycle4')
)
  group_by(cycle4,faaAge,gender) %>% summarise( N=100*Nc/Nt )

fmtWTLB( NN )
```

	cycle4	Years	faaAge	gender	N
1	4	2003-2006	Infant	None	2.9
2	4	2003-2006	Child	None	15.0
3	4	2003-2006	Adult	Male	39.7
4	4	2003-2006	Adult	Female	42.4
5	6	2007-2010	Infant	None	2.9
6	6	2007-2010	Child	None	14.8
7	6	2007-2010	Adult	Male	40.1
8	6	2007-2010	Adult	Female	42.2
9	8	2011-2014	Infant	None	2.4
10	8	2011-2014	Child	None	14.6
11	8	2011-2014	Adult	Male	40.4
12	8	2011-2014	Adult	Female	42.6
13	10	2015-2018	Infant	None	2.4
14	10	2015-2018	Child	None	13.7
15	10	2015-2018	Adult	Male	40.8
16	10	2015-2018	Adult	Female	43.1

From this data, calculating  $[W]$  values for use in operational systems is simplest with a hand calculator. For completeness, implementation of formulas (10), (11), and (12) are also shown in R.

```
library(tidyr) # for pivot

Wx <- left_join( Step5 %>% select(cycle4,faaAge,gender,W)
  pivot_wider(names_from=c(faaAge,gender),values_from=W,names_prefix='W_'),
  NN
  pivot_wider(names_from=c(faaAge,gender),values_from=N,names_prefix='N_'),
  by=c('cycle4') )
  rename( N_Infant=N_Infant_None, W_Infant=W_Infant_None )
  mutate(
    W_Adult_NHANES = ( N_Adult_Male * W_Adult_Male +
                      N_Adult_Female * W_Adult_Female
                    ) / ( N_Adult_Male + N_Adult_Female ),
    W_Adult_5050 = 0.5*( W_Adult_Male + W_Adult_Female ), # Formula 12
```

```

dw_Infant_NHANES = N_Infant / ( N_Adult_Female + N_Adult_Male ) * W_Infant,

'W_Adult_NHANES+Infant' = W_Adult_NHANES + dw_Infant_NHANES, # Formula 10
'W_Adult_5050+Infant'   = W_Adult_5050   + dw_Infant_NHANES, # Formula 11
)

fmtWTLB( Wx %>% select(cycle4,W_Adult_NHANES,W_Adult_5050,dw_Infant_NHANES) )

  cycle4      Years W_Adult_NHANES W_Adult_5050 dw_Infant_NHANES
1         4 2003-2006          170.5          171.0             0.8
2         6 2007-2010          171.6          171.9             0.7
3         8 2011-2014          172.7          173.0             0.6
4        10 2015-2018          175.3          175.6             0.6

fmtWTLB( Wx %>% select(cycle4,'W_Adult_NHANES+Infant','W_Adult_5050+Infant') )

  cycle4      Years W_Adult_NHANES+Infant W_Adult_5050+Infant
1         4 2003-2006             171.3             171.7
2         6 2007-2010             172.3             172.7
3         8 2011-2014             173.3             173.6
4        10 2015-2018             175.9             176.2

```

In the data shown, the weight increase resulting from a “50:50” M:F ratio compared to a ratio estimated from NHANES does not exceed half a pound, which is comparable in scale to rounding to whole pounds.

Anecdotally, fewer adults travel with infants than have infant children, so using the proportion of Infants to Adults in the NHANES population almost certainly over-estimates the body mass of Infants on typical flights. Operator-specific data is expected to reduce the magnitude of adjustments for infants, which is also comparable in scale to rounding to whole pounds.

In Part 3 of [METHOD DEFINITION](#), appropriate ages for crewmembers differ significantly from the definition of Adult in AC 120-27F. Since age requirements for crew are not universal, the definition uses typical values. For pilots the minimum comes from 14 CFR 61.153(a)(1) for an unrestricted airline transport pilot (ATP) certificate and the maximum from 15 CFR 121.383(d) for air carrier limitations on pilot age. For Flight Attendants, the minimum age to serve alcohol in any US state has been observed as a common hiring requirement and is used here.

Commands used above to implement Steps 1 through 5 are repeated here in a condensed format, using **P** for Pilots and **F** for Flight Attendants.

```

# Repeat Part 1, Steps 1 thru 5 using ages representative of Pilots
P0 <- AllData %>% filter( NotClothed & 23<=RIDAGEYR & 65>RIDAGEYR )
P1 <- mutate(P0, ww=WTLB*WTMEC4YR) %>%
  group_by(gender,cycle4) %>%
  summarise( ws=sum(ww)/sum(WTMEC4YR) )
P2 <- left_join(P0,P1,by=c('gender','cycle4')) %>%
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 ) %>%
  group_by(gender,cycle4,ws) %>%
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*ss/ws/sqrt(n) )

```



```

P3 <- left_join(P0,P2,by=c('gender','cycle4'))           %>%
  filter( 2 >= abs((WTLB-ws)/ss) )
P4 <- mutate(P3, wr=WTLB*WTMEC4YR)                       %>%
  group_by(gender,cycle4)                               %>%
  summarise( W=sum(wr)/sum(WTMEC4YR) )
P5 <- left_join(P3,P4,by=c('gender','cycle4'))           %>%
  mutate( wr=WTMEC4YR*(WTLB-W)^2 )                     %>%
  group_by(gender,cycle4,W)                             %>%
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*S/W/sqrt(n) )
fmtWTLB(P5 %>% select(cycle4,gender,n,W,S,'te%'))

```

	cycle4	Years	gender	n	W	S	te%
1	4	2003-2006	Male	2766	192.5	33.8	0.7
2	6	2007-2010	Male	3650	193.0	35.5	0.6
3	8	2011-2014	Male	3513	193.1	35.1	0.6
4	10	2015-2018	Male	3288	196.2	39.5	0.7
5	4	2003-2006	Female	3136	162.1	36.5	0.8
6	6	2007-2010	Female	3893	162.8	36.2	0.7
7	8	2011-2014	Female	3709	165.9	37.4	0.7
8	10	2015-2018	Female	3631	167.2	38.9	0.8

# Repeat Part 1, Steps 1 thru 5 using ages representative of Flight Attendants

```

F0 <- AllData %>% filter( NotClothed & 21<=RIDAGEYR )
F1 <- mutate(F0, ww=WTLB*WTMEC4YR)                       %>%
  group_by(gender,cycle4)                               %>%
  summarise( ws=sum(ww)/sum(WTMEC4YR) )
F2 <- left_join(F0,F1,by=c('gender','cycle4'))           %>%
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 )                     %>%
  group_by(gender,cycle4,ws)                             %>%
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*ss/ws/sqrt(n) )
F3 <- left_join(F0,F2,by=c('gender','cycle4'))           %>%
  filter( 2 >= abs((WTLB-ws)/ss) )
F4 <- mutate(F3, wr=WTLB*WTMEC4YR)                       %>%
  group_by(gender,cycle4)                               %>%
  summarise( W=sum(wr)/sum(WTMEC4YR) )
F5 <- left_join(F3,F4,by=c('gender','cycle4'))           %>%
  mutate( wr=WTMEC4YR*(WTLB-W)^2 )                     %>%
  group_by(gender,cycle4,W)                             %>%
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*S/W/sqrt(n) )
fmtWTLB(F5 %>% select(cycle4,gender,n,W,S,'te%'))

```

	cycle4	Years	gender	n	W	S	te%
1	4	2003-2006	Male	4080	190.6	34.2	0.6
2	6	2007-2010	Male	5133	191.3	35.6	0.5
3	8	2011-2014	Male	4724	191.1	35.0	0.5
4	10	2015-2018	Male	4626	194.5	38.7	0.6
5	4	2003-2006	Female	4400	159.3	35.0	0.6

6	6	2007-2010	Female	5403	160.6	35.2	0.6
7	8	2011-2014	Female	4986	162.8	35.9	0.6
8	10	2015-2018	Female	4928	164.7	37.1	0.6

Generally, ratios for  $M_P:F_P$  and  $M_{FA}:F_{FA}$  may be estimated from an operator’s total employed crewmembers. Sample calculations using 50:50 and 99:1 are shown for reference when implementing formula (13) with no claim that these ratios are applicable to any specific operation.

# for example only, apply sample M:F ratios for crewmembers using Formula (13)  
`F13 <- function(M,F,W_M,W_F) { ( M*W_M + F*W_F )/(M+F) }`

```
fmtWTLB( P5 %>% select(cycle4,gender,W) %>%
  pivot_wider(names_from=c(gender),values_from=W,names_prefix='W_P__') %>%
  mutate( W_P_50=F13( 50,50, W_P__Male,W_P__Female ),
    W_P_99=F13( 99, 1, W_P__Male,W_P__Female )) )
```

	cycle4	Years	W_P__Male	W_P__Female	W_P_50	W_P_99
1	4	2003-2006	192.5	162.1	177.3	192.2
2	6	2007-2010	193.0	162.8	177.9	192.7
3	8	2011-2014	193.1	165.9	179.5	192.8
4	10	2015-2018	196.2	167.2	181.7	195.9

```
fmtWTLB( F5 %>% select(cycle4,gender,W) %>%
  pivot_wider(names_from=c(gender),values_from=W,names_prefix='W_FA_') %>%
  mutate( W_FA_50=F13( 50,50, W_FA_Male,W_FA_Female ),
    W_FA_99=F13( 99, 1, W_FA_Male,W_FA_Female )) )
```

	cycle4	Years	W_FA_Male	W_FA_Female	W_FA_50	W_FA_99
1	4	2003-2006	190.6	159.3	175.0	190.3
2	6	2007-2010	191.3	160.6	175.9	190.9
3	8	2011-2014	191.1	162.8	176.9	190.8
4	10	2015-2018	194.5	164.7	179.6	194.2

**SUMMARY VALUES**

The following tables show results of the METHOD DEFINITION applied to NHANES 4-year cycles from 2003 to 2018.

These values were calculated using a Microsoft Excel spreadsheet and validated using the R scripts in DISCUSSION. To the number of decimal places reported, there is no difference between calculation tools used.

NHANES 4 Yr. Cycle	Adult Male			Adult Female			Child			Infant		
	$n_f$	$W_M$	$\sigma$	$n_f$	$W_F$	$\sigma$	$n_f$	$W_C$	$\sigma$	$n_f$	$W_I$	$\sigma$
2003-2006	5940	186.1	36.1	6321	155.9	35.1	4087	62.0	26.4	1494	21.8	5.1
2007-2010	6291	186.5	37.3	6435	157.3	35.3	4156	61.5	26.7	1382	21.1	4.9
2011-2014	5879	186.8	37.1	6125	159.3	36.4	4303	62.4	29.9	1167	20.9	5.2
2015-2018	5689	189.6	40.5	5937	161.6	37.2	3692	62.9	26.3	1118	21.1	5.0

Table 4: Survey-derived Average Crewmember Weights from NHANES Data (LB)												
NHANES 4 Yr. Cycle	Pilot, Male			Pilot, Female			Flight Attendant, Male			Flight Attendant, Female		
	$n_f$	$W_{P,M}$	$\sigma$	$n_f$	$W_{P,F}$	$\sigma$	$n_f$	$W_{FA,M}$	$\sigma$	$n_f$	$W_{FA,F}$	$\sigma$
2003-2006	2766	192.5	33.8	3136	162.1	36.5	4080	190.6	34.2	4400	159.3	35.0
2007-2010	3650	193.0	35.5	3893	162.8	36.2	5133	191.3	35.6	5403	160.6	35.2
2011-2014	3513	193.1	35.1	3709	165.9	37.4	4724	191.1	35.0	4986	162.8	35.9
2015-2018	3288	196.2	39.5	3631	167.2	38.9	4626	194.5	38.7	4928	164.7	37.1

Table 5: Adjusted Body Weights for Sample M:F Ratios (LB)								
NHANES 4 Yr. Cycle	[ $W_x$ ] Not Adjusted for Infants					[ $W_x$ ] Adjusted for Infants		
	Pilots		Flight Attendants		Adults	Adults	Adults	Adults
M:F	50:50	99:1	50:50	99:1	NHANES	50:50	NHANES	50:50
2003-2006	177.3	192.2	175.0	190.3	170.5	171.0	171.3	171.7
2007-2010	177.9	192.7	175.9	190.9	171.6	171.9	172.3	172.7
2011-2014	179.5	192.8	176.9	190.8	172.7	173.0	173.3	173.6
2015-2018	181.7	195.9	179.6	194.2	175.3	175.6	175.9	176.2

#### REFERENCES

Centers for Disease Control and Prevention (CDC). National Center for Health Statistics (NCHS). National Health and Nutrition Examination Survey Data. Hyattsville, MD: U.S. Department of Health and Human Services, Centers for Disease Control and Prevention, 2003-2018, [www.cdc.gov/nchs/nhanes](http://www.cdc.gov/nchs/nhanes).

Chen TC, Clark J, Riddles MK, Mohadjer LK, Fakhouri THI. National Health and Nutrition Examination Survey, 2015–2018: Sample design and estimation procedures. National Center for Health Statistics. Vital Health Stat 2(184). 2020.

Federal Aviation Administration, Advisory Circular 120-27 “Aircraft Weight and Balance Control”: (original) 1968; 120-27C 1995; 120-27D 2004; 120-27E 2005; 120-27F 2019.

Fryar CD, Kruszon-Moran D, Gu Q, Ogden CL. Mean body weight, height, waist circumference, and body mass index among adults: United States, 1999–2000 through 2015–2016. National Health Statistics Reports; no 122. Hyattsville, MD: National Center for Health Statistics. 2018. <https://www.cdc.gov/nchs/data/nhsr/nhsr122-508.pdf>

Lumley, Thomas. Complex sampling and R. University of Washington Biostatistics and “user” Conference, Rennes, France. 2009. <http://faculty.washington.edu/tlumley/tutorials/survey-user.pdf>

National Health and Nutrition Examination Survey (NHANES): Anthropometry Procedures Manual, January 2017.

National Health and Nutrition Examination Survey: Analytic Guidelines, 2011-2014 and 2015-2016 December 14, 2018.

## APPENDIX A

To run this script, those new to R will need to install the language core, which is freely available for a variety of platforms from [www.r-project.org](http://www.r-project.org). Use is simplified by installing a graphical development interface, such as RStudio Desktop (available as of writing at no cost from [rstudio.com](https://www.rstudio.com/)), and pasting the entire code into it directly. The code may be executed line by line or by selecting "Run All".

Packages required for installation are: `dplyr`, `foreign`, `survey`, and `tidyr`.

Other modifications are generally not required except to assure the path at `setwd(...)` exists and is accessible for writing. System specific details such as what directories may be accessible to the user are not addressed here.

A small font and large page are used to assist pasting from this document into plain text.

#-----

```
# Script for DISCUSSION section 'NHANES USE FOR STANDARD AVERAGE BODY WEIGHT'
# (c) 2020 Society of Aircraft Performance and Operations Engineers (SAPOE)
#
# SAPOE is a member-based organization promoting the safety and efficiency of
# flight through knowledge of aircraft performance and weight and balance
# principles.
#
# Except for brief quotations with appropriate citation, copies may be made
# and distributed only of the complete document including cover page and
# this disclaimer.
#
# SAPOE, the authors, and reviewers assume no liability whatsoever and make
# no warranty of any kind. Anyone who uses this report, or the source code
# and data contained in it, is solely responsible for their own operations
# and for any outcome of such use.
```

```
# The 'dplyr' library simplifies and improves readability, especially with
# 'mutate()' to add calculated columns to a data table,
# 'group_by()' to define categories within the data, and
# 'summarise()' [note spelling] to calculate summary statistics
library(dplyr)
```

```
# Define a helper to keep cycle number and year conventions consistent
cycleYears <- function( cycle2or4, cycleNumbers )
{
  yr <- 2000+2*(cycleNumbers-1)
  sprintf('%s-%s',yr-(cycle2or4-1),yr)
}
```

```
# Define a helper to download needed data files to the working directory
downloadDataCycles <- function( cycleNumbers )
{
  for( cn in cycleNumbers ){
    files <- sapply(c( 'DEMO_%s.XPT','BMX_%s.XPT' ),sprintf,LETTERS[cn])
    for( f in files ){
      download.file(
        sprintf('https://wwwn.cdc.gov/nchs/nhanes/%s/%s',cycleYears(2,cn),f),
        f, mode='wb'
      )
    }
  }
}
```

```
# Change the working directory to one the current user can save files locally
setwd( 'C:/NHANES/' )
```

```
# Download XPT files for the cycles of interest
# Do this just once (ever) and then comment out this line
downloadDataCycles(3:10)
```

```
# The 'foreign' library reads SAS Transport(XPT) files
library(foreign)
```

```
# Define a helper to load the needed data columns files in the working directory
importDataCycles <- function( cycleNumbers )
{
```

```
  allData <- data.frame(NULL)
  for( cn in cycleNumbers ){
    allData <- bind_rows(allData,left_join(
      subset(
        select(
          read.xport(sprintf('DEMO_%s.XPT', LETTERS[cn])),
          # only needed columns from the DEMO file here
          'SEQN','SDDSRVYR','RIDAGEYR','RIAGENDR','RIDEXPRG',
          'SDMVSTRA','SDMVPSU','WTMEC2YR'
        ),
        list.na(WTMEC2YR)), # exclude participants without an exam record
      select(
        read.xport(sprintf('BMX_%s.XPT', LETTERS[cn])),
        # only needed columns from the BMX file here
        'SEQN','BMXWT','BMIWT'
      ),
      by = 'SEQN'
    )
  )
  }
  return(allData)
}
```

```
# Load the data into memory once each session
RawData <- importDataCycles(3:10)
```

```
# Create a copy with calculated columns added (preserve RawData)
# Don't remove any rows, or survey design parameters are lost.
AllData <- mutate(RawData,
  # 'one' is needed to count records
  one = 1,
```

```
# combine 2-year cycles as recommended by Analytical Guidelines
```

```
  cycle4 = 2*ceiling(SDDSRVYR/2),
  WTMEC4YR = WTMEC2YR/2,

  # an alias to ease comparing 4- and 2-year cycles
  cycle2 = SDDSRVYR,
  # use with WTMEC2YR

  # make it easy to work in US units
  WTLB = BMXWT / 0.45359237, # kg to lb

  # use cut(right=FALSE) to build age intervals that include the lower
  # and exclude the upper boundary
  #
  # for FAA, Infants have not yet reached 2nd birthday.
  # Children have not yet reached 13th birthday.
  faaAge = cut(RIDAGEYR, breaks=c(-Inf, 2, 13, Inf),
    right=FALSE, labels=c( 'Infant', 'Child', 'Adult' )),
  gender=factor(ifelse(RIDAGEYR>12,RIAGENDR,0),labels=c('None','Male','Female')),
  #
  # for comparison to CDC/NHSR, Adults are age 20 and over
  cdcAge = cut(RIDAGEYR, breaks=c(-Inf, 20, Inf),
    right=FALSE, labels=c( 'Child', 'Adult'  ))
)

# Add columns used to filter the data, like 'inAnalysis' in tutorials
# Don't remove any rows, or survey design parameters are lost.
AllData <- mutate(AllData,
  NotMissing = ( !is.na(BMXWT) ),
  NotClothed = ( !is.na(BMXWT) & (is.na(BMIWT) | 3!=BMIWT) ),
  NotPregnant = ( !is.na(BMXWT) & (is.na(RIDEXPRG) | 1!=RIDEXPRG) )
)
# for RIDEXPRG, 1=Yes, 2=No, 3=Unknown, Missing => No

# based heavily on https://wwwn.cdc.gov/nchs/nhanes/tutorials/samplecode.aspx
# and https://wwwn.cdc.gov/nchs/data/tutorials/DB303_Fig1_R.R
library(survey)

# Define survey design parameters for overall dataset
NHANES2 <- svydesign(data=AllData,
  id=~SDMVPSU, strata=~SDMVSTRA, weights=~WTMEC2YR, nest=TRUE)
NHANES4 <- svydesign(data=AllData,
  id=~SDMVPSU, strata=~SDMVSTRA, weights=~WTMEC4YR, nest=TRUE)

# Define a function to get the weighted mean and standard error estimate
# using survey design parameters and join to the unweighted count
svyWTLB <- function(design, byFormula)
{
  # Get mean, stderr, and unweighted sample size
  c <- svyby( ~WTLB, byFormula, design, unwt.d.count )
  p <- svyby( ~WTLB, byFormula, design, svy.mean )
  # select(c,-se) excludes the standard error of the count
  # suppressMessages() just hides output of the implicit "by" columns
  suppressMessages( outData <- left_join(select(c,-se), p) )
  return(outData)
}

# Define a function to get consistent output formatting
fmtWTLB <- function(cycleData)
{
  c2 <- match("cycle2",names(cycleData))
  if(!is.na(c2)) cycleData <- mutate(cycleData,
    Years=cycleYears(2,cycle2), .after=cycle2)
  c4 <- match("cycle4",names(cycleData))
  if(!is.na(c4)) cycleData <- mutate(cycleData,
    Years=cycleYears(4,cycle4), .after=cycle4)
  print.data.frame( cycleData %>% mutate_if(is.numeric,round,digits=1) )
}

# confirm definitions above here are correct by comparing to
# https://www.cdc.gov/nchs/data/nhsr/nhsr122-508.pdf

fmtWTLB( svyWTLB( subset( NHANES2, NotPregnant ), ~cycle2+cdcAge+gender ) ) %>%
  filter(cdcAge=='Adult'&cycle2<10) # NHSR Table 2 has Adults only

# run the same formula with faaAge, and not excluding pregnant females
#fmtWTLB( svyWTLB( subset( NHANES2, NotPregnant ), ~cycle2+cdcAge+gender ) )
fmtWTLB( svyWTLB( subset( NHANES4, NotClothed ), ~cycle4+faaAge+gender ) )

# A copy of the data excluding clothed sample participants
Step0 <- AllData %>% filter(NotClothed)

# Calculate a raw weighted mean, w-bar-star (ws) using Formula (1)
Step1 <- mutate(Step0, ww=WTLB*WTMEC4YR) %>%
  group_by(faaAge,gender,cycle4) %>%
  summarise( ws=sum(ww)/sum(WTMEC4YR) )

# Sum weighted residuals (wr) for each row to get sigma-star (ss) using Formula (2)
Step2 <- left_join(Step0,Step1,by=c('faaAge','gender','cycle4')) %>%
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 ) %>%
  group_by(faaAge,gender,cycle4,ws) %>%
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*ss/ws/sqrt(n) )

fmtWTLB(Step2 %>% select(cycle4,faaAge,gender,n,ws,ss,'te%'))

# Filter by excluding values more than twice ss from the raw mean
Step3 <- left_join(Step0,Step2,by=c('faaAge','gender','cycle4')) %>%
  filter( 2 >= abs((WTLB-ws)/ss) )

# Calculate a weighted mean, W using Formula (3)
Step4 <- mutate(Step3, wr=WTLB*WTMEC4YR) %>%
  group_by(faaAge,gender,cycle4) %>%
  summarise( W=sum(wr)/sum(WTMEC4YR) )
```

```
# Sum weighted residuals (wr) for each row to get sigma (S) using Formula (4)
Step5 <- left_join(Step3,Step4,by=c('faaAge','gender','cycle4')) %>%
  mutate( wr=WTMEC4YR*(WTLB-W)^2 ) %>%
  group_by(faaAge,gender,cycle4,W) %>%
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
  mutate( 'te%'=196*S/W/sqrt(n) )
```

```
fmtWTLB(Step5 %>% select(cycle4,faaAge,gender,n,W,S,'te%'))
```

```
# Calculate population fractions using Formula (7)
NN <- left_join(
  Step3 %>% group_by(cycle4,faaAge,gender) %>% summarise( Nc=sum(WTMEC4YR) ),
  Step5 %>% group_by(cycle4, ) %>% summarise( Nt=sum(WTMEC4YR) ),
  by=c('cycle4')
) %>%
  group_by(cycle4,faaAge,gender) %>% summarise( N=100*Nc/Nt )
```

```
fmtWTLB( NN )
```

```
library(tidyr) # for pivot
```

```
Wx <- left_join( Step5 %>% select(cycle4,faaAge,gender,W) %>%
  pivot_wider(names_from=c(faaAge,gender),values_from=W,names_prefix='W_'),
  NN
) %>%
  pivot_wider(names_from=c(faaAge,gender),values_from=N,names_prefix='N_'),
  by=c('cycle4') ) %>%
  rename( N_Infant=N_Infant_None, W_Infant=W_Infant_None ) %>%
  mutate(
    W_Adult_NHANES = ( N_Adult_Male * W_Adult_Male +
      N_Adult_Female * W_Adult_Female
    ) / ( N_Adult_Male + N_Adult_Female ),
    W_Adult_5050 = 0.5*( W_Adult_Male + W_Adult_Female ), # Formula 12
```

```
  dW_Infant_NHANES = N_Infant / ( N_Adult_Female + N_Adult_Male ) * W_Infant,

  'W_Adult_NHANES+Infant' = W_Adult_NHANES + dW_Infant_NHANES, # Formula 10
  'W_Adult_5050+Infant' = W_Adult_5050 + dW_Infant_NHANES, # Formula 11
)
```

```
fmtWTLB( Wx %>% select(cycle4,W_Adult_NHANES,W_Adult_5050,dW_Infant_NHANES) )
```

```
fmtWTLB( Wx %>% select(cycle4,'W_Adult_NHANES+Infant','W_Adult_5050+Infant') )
```

```
# Repeat Part 1, Steps 1 thru 5 using ages representative of Pilots
```

```
P0 <- AllData %>% filter( NotClothed & 23<=RIDAGEYR & 65>RIDAGEYR )
```

```
P1 <- mutate(P0, ww=WTLB*WTMEC4YR) %>%
```

```
  group_by(gender,cycle4) %>%
```

```
  summarise( ws=sum(ww)/sum(WTMEC4YR) )
```

```
P2 <- left_join(P0,P1,by=c('gender','cycle4')) %>%
```

```
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 ) %>%
```

```
  group_by(gender,cycle4,ws) %>%
```

```
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
```

```
  mutate( 'te%'=196*ss/ws/sqrt(n) )
```

```
P3 <- left_join(P0,P2,by=c('gender','cycle4')) %>%
```

```
  filter( 2 >= abs((WTLB-ws)/ss) )
```

```
P4 <- mutate(P3, wr=WTLB*WTMEC4YR) %>%
```

```
  group_by(gender,cycle4) %>%
```

```
  summarise( W=sum(wr)/sum(WTMEC4YR) )
```

```
P5 <- left_join(P3,P4,by=c('gender','cycle4')) %>%
```

```
  mutate( wr=WTMEC4YR*(WTLB-W)^2 ) %>%
```

```
  group_by(gender,cycle4,W) %>%
```

```
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
```

```
  mutate( 'te%'=196*S/W/sqrt(n) )
```

```
fmtWTLB(P5 %>% select(cycle4,gender,n,W,S,'te%'))
```

```
# Repeat Part 1, Steps 1 thru 5 using ages representative of Flight Attendants
```

```
F0 <- AllData %>% filter( NotClothed & 21<=RIDAGEYR )
```

```
F1 <- mutate(F0, ww=WTLB*WTMEC4YR) %>%
```

```
  group_by(gender,cycle4) %>%
```

```
  summarise( ws=sum(ww)/sum(WTMEC4YR) )
```

```
F2 <- left_join(F0,F1,by=c('gender','cycle4')) %>%
```

```
  mutate( wr=WTMEC4YR*(WTLB-ws)^2 ) %>%
```

```
  group_by(gender,cycle4,ws) %>%
```

```
  summarise( n=sum(one), ss=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
```

```
  mutate( 'te%'=196*ss/ws/sqrt(n) )
```

```
F3 <- left_join(F0,F2,by=c('gender','cycle4')) %>%
```

```
  filter( 2 >= abs((WTLB-ws)/ss) )
```

```
F4 <- mutate(F3, wr=WTLB*WTMEC4YR) %>%
```

```
  group_by(gender,cycle4) %>%
```

```
  summarise( W=sum(wr)/sum(WTMEC4YR) )
```

```
F5 <- left_join(F3,F4,by=c('gender','cycle4')) %>%
```

```
  mutate( wr=WTMEC4YR*(WTLB-W)^2 ) %>%
```

```
  group_by(gender,cycle4,W) %>%
```

```
  summarise( n=sum(one), S=sqrt(sum(wr)/((n-1)/n*sum(WTMEC4YR))) ) %>%
```

```
  mutate( 'te%'=196*S/W/sqrt(n) )
```

```
fmtWTLB(F5 %>% select(cycle4,gender,n,W,S,'te%'))
```

```
# for example only, apply sample M:F ratios for crewmembers using Formula (13)
```

```
F13 <- function(M,F,W_M,W_F) { ( M*W_M + F*W_F )/(M+F) }
```

```
fmtWTLB( P5 %>% select(cycle4,gender,W) %>%
```

```
  pivot_wider(names_from=c(gender),values_from=W,names_prefix='W_P_') %>%
```

```
  mutate( W_PA_50=F13( 50,50, W_PA_Male,W_PA_Female ),
```

```
    W_PA_99=F13( 99, 1, W_PA_Male,W_PA_Female ) )
```

```
#-----
```

END