

Simulation of Clinical Trials Workshop Supplementary Material

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Code Segments Example 1—Bioequivalence Example with Sample Size Re-Estimation in SAS

```
%let NumIter = 1000;
%let TargetN = 240;
%let MaxN = 300;

data phasel;
input CV DORate InterimPoint;
do iter = 1 to &NumIter;
  do patient=1 to &MaxN;
    day = floor((patient-1)/6) + 1;
    trt = mod(patient-1, 3) + 1;
    if patient<=&TargetN then PlannedGroup=1; else PlannedGroup=0;
    if patient<=&TargetN*InterimPoint then ForInterim=1; else ForInterim=0;
    logCmax = normal(0)*CV;
    if uniform(0) <= DORate then DO=1; else DO=0;
    if trt=1 then logCmax = logCmax + 0.05; ** trt=2 and 3 are the reference products **;
    output;
  end;
end;
cards;
0.33 0 0.333
0.33 0.25 0.333
Etc.
;
proc sql;
  create table BlindedInterimResults as
  select CV, DORate, InterimPoint, iter, N(DO) as NsoFar, mean(DO) as DORateObs,
         std(logCmax) as CVhat, max(lastPK)+&PKanalysisTime+&AssayTime as
StartAdditionalCohort
  from phasel
  where ForInterim=1
  group by CV, DORate, InterimPoint, iter;
quit;
data Predicted;
set BlindedInterimResults;
NumEvaluable = (1-DORateObs)*&Cohort*&NumCohort/3;
if NumEvaluable < 60 then MaxN=10; else MaxN=8;

*** Calculate Sample Size ***;
```

```

CVpt1 = 0.25;      CVpt2 = 0.275;    CVpt3 = 0.3;
SS1 = 36;    SS2 = 43;    SS3 = 50;
array CVpt {*} CVpt1-CVpt3;
array SS {*} SS1-SS3;

if CVhat <= CVpt1 then SSarm = SS1;
else
    do i=2 to 12;
        if CVpt{i-1} < CVhat <= CVpt{i} then SSarm = SS{i};
    end;
Recruit = SSarm/(1-DORateObs);    ** Per arm sample size **;
Recruit = 3*Recruit;              ** Sample size total **;
Recruit = ceil(max(NsoFar, Recruit));    ** Keep what we already have **;

keep CV DORate InterimPoint iter MaxN SSarm Recruit StartAdditionalCohort NsoFar;
run;
proc sort data=phase1;
    by CV DORate InterimPoint iter;
run;
data FinalData;
    merge phase1 Predicted AdditionalCohortsDates MaxPlannedGroup;
    by CV DORate InterimPoint iter;
    if patient <= max(NsoFar, Recruit);
run;
proc sort data=FinalData;
    by CV DORate InterimPoint iter;
run;

ods listing close;
ods output Estimates=estimates;
proc mixed data=FinalData(where=(DO=0));
    by CV DORate InterimPoint iter;
    class trt;
    model logCmax = trt;
    estimate 'TrtDiff 1 vs 2' trt 1 -1 0 / cl alpha=0.1;
    estimate 'TrtDiff 1 vs 3' trt 1 0 -1 / cl alpha=0.1;
    estimate 'TrtDiff 2 vs 3' trt 0 1 -1 / cl alpha=0.1;
run;
ods listing;
data CIinLimits;
    set estimates;

```

```
    if log(0.8) <= lower and upper <= log(1.25) then InLimits=1;
    else InLimits=0;
run;
proc sql;
    create table TrialResults as
        select CV, DOrate, InterimPoint, iter, max(InLimits) as AnyIn, min(InLimits) as AllIn
        from CIinLimits
        group by CV, DOrate, InterimPoint, iter;

quit;
```

Code Segments Example 2—Time Response and Dose Response Model in R

Selected code segments. Not entire program.

Create responses for each subject

```
subject.data <- function(dose=300, dose.effect)
{
  ##### Generates data based on the linear model from the Phase I study
  #####
  ##### Inputs:
  #####      dose.effect = 0 (no effect), 1 (Phase I model), 2 (strong, 15 characters)
  #####
  ##### Outputs:
  #####      data.frame with columns time, val
  #####      Time points are 4, 8, 12, 16, 24 weeks

  #####
  ### From lm() output on Phase I data
  ##
  ##Coefficients:
  ##              Estimate Std. Error t value Pr(>|t|)
  ##(Intercept)      xxxxxx  <In actual program had the  output>
  ##week:Dose        xxxxxxxx
  ##week:as.factor(volume)1  xxxxxx
  ##
  #####

  weeks <- c(4, 8, 12, 16, 24)
  n <- length(weeks)

  eta <- rnorm(1, 0, 10)
  epsilon <- rnorm(n, 0, 10)
```

```

if(dose==0)
{
    val <- 0          - xxxxxx*dose*dose.effect + eta + epsilon
}
else
{
    val <- 0          + eta + epsilon
}

result <- data.frame(week=weeks, val=val)
result
}

```

Run Replicated Trials with Parameters Provided

```

design <- function(iter=1, sample.size, dose.effect, alpha=0.05, design)
{
    run.trial <- function(sample.size, dose.effect, design)
    {
        des2 <- design[design$status == 1,]
        DesignPts <- dim(des2)[1]
        NumRep <- trunc( sample.size/DesignPts ) + 1
        doses <- rep(design$dose1, sample.size)[1:sample.size]

        data.out <- data.frame(
            dose1=doses,
            Response=rep(0,length(doses)),
            dose=as.factor(doses)
        )

        for(i in 1:dim(data.out)[1])
        {
            subj <- subject.data(dose=data.out[i,"dose1"], dose.effect)
            row <- subj$week==24

```

Simulate trial with patients randomized to various doses

Number of replicates; will want to start with just a few for debugging, and then set to a large number (e.g., 10,000) to compute results

```

        data.out[i,"VAS"] <- subj[row,3]
    }
    trial.lm <- lm(VAS~dose:volume-1, data=data.out)
    pvalues <- summary(trial.lm)$coefficients[,4]
    pvalues[summary(trial.lm)$coefficients[,1]>0] <- 1
    pvalues
}

found.best <- function(trial.out, alpha, dose.effect, volume.effect, design)
{
    DV <- function(labels)
    {
        N <- length(labels)
        doses <- rep(0, N)
        vols <- rep(0, N)
        DVsplit <- strsplit(labels, ":")
        for(i in 1:N)
        {
            doses[i] <- as.numeric(strsplit(DVsplit[[i]][1], "dose")[[1]][2])
        }
        data.frame(dose=doses)
    }

    DVoptions <- DV(names(trial.out))
    if(dose.effect==0)
    {
        val <- min(trial.out)>alpha
    }
    else
    {
        ##### Where is maximum effect #####
        max.effect <- trial.out == min(trial.out)

        ##### Where is the true maximum effect? #####
        max.dose <- max(DVoptions["dose"])
        is.best <- DVoptions$dose==max.dose
    }
}

```

We need to answer
the question: Did
we find the best
dose?

```

        val <- any(max.effect & is.best)
    }
    val
}

##### Start computations #####
K <- sum(design$status)
test.results <- matrix(rep(NA, (K+2)*iter), iter, K+2)

for(i in 1:iter)
{
    trial.out <- run.trial(sample.size, dose.effect, design)
    test.results[i, 1:K] <- trial.out
    test.results[i, K+1] <- min(trial.out)
    test.results[i, K+2] <- found.best(trial.out, alpha, dose.effect, design)
}

##### Summarize trial results #####
power <- rep(0, K+2)
Names <- c(names(trial.out), "overall", "best.dose")
names(power) <- Names
for(i in 1:(K+1))
{
    power[i] <- mean(ifelse(test.results[,i]<alpha,1,0))
}
power[K+2] <- mean(test.results[,K+2])
power
}

```

Run the trials repeatedly, so we can summarize operating characteristics

Calculate power

Run Replicated Trials with Parameters Provided

```

sim <- function(iter=1, dose.cond=c(0, 1, 2, 0, 1, 2), alpha=0.05)
{
    ##### Set everything up
    fixed.power <- function(output)
    {

```



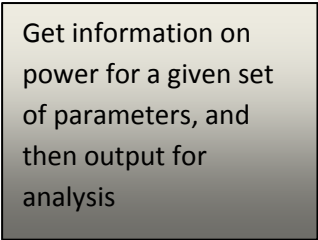
```

        output[(length(output)-2):length(output)]
    }

    K <- length(dose.cond)
    power.curve <- data.frame(P560=rep(0,K), overall=rep(0, K),
                             best.dose=rep(0, K))

    ##### Vary the parameters, and call the simulation function
    for(i in 1:K)
    {
        fixed.output <- design(iter, sample.size=180,
                               dose.effect=dose.cond[i],
                               alpha=0.05, make.design2())
        power.curve[i,] <- fixed.power(fixed.output)
    }
    cbind(data.frame(n=rep(iter, K), dose=dose.cond), power.curve)
}

```



Get information on power for a given set of parameters, and then output for analysis