

hello Im Rob bees and Im faculty at
Buffalo and uh Im uh presenting today
on a
complimentary uh vignette to the lecture
on population
pharmacokinetics uh these are broken
into multiple parts and this is part one
part one focus on exploratory data
analysis and the initial elements that
are necessary prior to to undertaking a
formal population pharmacokinetic
analysis so theres some general steps
in a population pharmacokinetic modeling
project the first of course is to
collect the data so some design some
experiment is designed or question is is
is arrived at an experiment is designed
data are generated and then we collect
the
data uh we need you know one needs to
understand exactly how the data were
generated uh theres data formatting
involved depending on what software or
what approach you might use for the
analysis uh theres an exploratory data

analysis step and then model build

approaches a general schematic is shown

in this slide with the general steps for
the population pharmacokinetic modeling
project and our part one of our vignette
really focuses on this top part here the
research question experiments data and
visualization data of data with a
particular focus on the nature of the
data and the visualization of the data
before we get to things such as model
context to to explore that will be taken
up in the
subsequent uh
vignette again were going to use
actually a case study with copine and
these are kipine concentration
measurements that were used for
population pharmacokinetic analysis for
data that were generated from the
clinical antic psychotic Trials of
intervention Effectiveness this is a an
imh funded study that
completed
uh over over a decade
ago um were going to again familiar

ourselves with the study design and
understand the data generating process
were going through going to go through
a very basic uh illustration of how to
construct a data set for population
pharmacokinetic analysis in this case a
nonmem ready data set nonmem is a one
of the many software available to
conduct this type of
analysis uh and then uh uh we will go
through an exploratory data analysis
evaluating the different characteristics
of the data that were
collected uh this is done prior to the
formal model building process and is
often done with a data wizard uh or with
a statistical programming language such
as

R uh things that we will be evaluating
are the numbers of observations numbers
of individ in this case concentration
observations
uh number of individuals what were the
doses administered what were what are
the patient specific
characteristics what is the distribution

of times after dose for the
concentration measurements and then any
information that one can glean from a
typical population pharmacokinetic study
with respect to the concentration versus
time profiles both at the individual and
aggregate
levels uh this uh is important in terms
of evaluating uh those profiles at the
level of patient specific or study
specific
characteristics another part part two of
the
vignette uh will cover the population
pharmacokinetic model building and
Analysis strategy and is outside the
scope of this initial
vignette so a key step is the
familiarization with the data generating
process or the study
design when considering the study design
one should evaluate whether or not this
was an observational or a randomized
study this will
impact uh the consideration of potential
confounders was a study conducted in a

highly controlled setting like a
clinical research center what was the
concentration sampling Paradigm was it
random was it
optimized what would what was the dosage
Paradigm what was the frequency of doses
what were the range of doses
what covariants were collected IE what
patient specific characteristics or
demographics do we have in the data
set and then what was the nature of the
sample collection were biomarkers
collected in addition to concentration
was the timing recorded
appropriately so lets move on to the
construction of the data
set so when one is assembling data from
a study such as Katie study one may get
data in
multiple uh table formats here I have a
demographics
table uh showing weight um and sex by
study
subject I have a table of doses so you
have the study subject you see there are
multiple doses listed and the time of

those doses in
hours and then I have a table of
observations so here we have the study
subject the times the plasma
concentration
measured so you notice the study subject
is the linking variable
here and for a population analysis the
study subject that is the that is the
basic level of analysis one considers
that as sort of the basic unit of
analysis respect to the determination of
parameters
and so if we look at a typical nonmem
data set it is actually assembled from
those components you see the study
subject ID appears at the far
left and uh different uh elements here
but we will go through these uh
individually in a
moment Now sort of superimposed the
original table formats or at least
excerpts or subsets of those initial
table formats to illustrate this
see that our table that indicate study
subject and the dosage histories are

captured in the overall analysis data
set as the study subject ID here labeled
pound ID and you can see these are
listed
here the dose in milligrams in this case
appears in the AMT or amount column and
you can see that they are noncontiguous
rows for most part because there are
observations down in
between and so this is now
sequenced by
time similarly the next table showing
steady subject time and plasma
concentration appear again in the ID
column time and the DV which in a nonm
parametric stands for dependent variables
this this is your observation this is
your
concentration and you can see now that
we've got we've got we've got plasma
concentrations interspersed here so the
nonm data set runs sequentially so you
go you have a
dose and you have an observation and it
runs sequentially in time time must be
ascending the other thing to note is

that we don't have any less than symbols
and we don't have the negative timing
typically you do not want to avoid
putting in a negative time off the newer
versions of nonm can manage that U and
you only
numerical uh elements can be present in
the data so you cannot have a less
than symbol you cannot have Alpha
Characters uh in the data set that's
mainly because this software uses
Fortran but even other analytical
software for the numerical components
will require numerical
inputs finally we have the demographic
table so again study subject right so
this is indexing
here weight you see weight appears and
weight is numerical and then we
have sex sex is listed in the original
table as female and
male but has to be
converted to a numerical flag in order
for this to be incorporated into the
analysis and so you see we have a column
we call were calling sex

M uh and zero is female one is male and
you can see these are the first two
individuals are those is female and
these
values are noted on every single Row for
those
individuals so now that weve put
together a general uh data
set there are a couple of other
components here that were
Incorporated these are the missing
dependent variable and the vent ID and
these are nonmem specific indicators
and Beyond the
scope uh of of the presentation
today but I will just make a quick note
that at least the missing dependent
variable is used to indicate when
theres an input process when theres
not an observation so here when you give
a dose you have a missing dependent
variable because you cannot have a dose
and an observation an input and an
output in the same
row um and so youll notice the missing
dependent variable is zero would you

have a plasma concentration this is

illustrated

here so let's move on to the exploratory
data analysis component we thought about

the study

design we formatted the data set so

were ready to start potentially doing

some some

analyses and now we're going to do an

exploratory data

analysis evaluating the information we

have in this data set prior to

modeling we're going to consider the
available observations and demographics

the covariate distributions the

correlations the signal ranges how

broad those signal ranges are they

something we want to test because we
have enough variety enough heterogeneity

in those measures to be able to pick up

a difference if there is

one going to evaluate the concentration

measurements the units the range of
observed concentrations any correlations

with covariates doses administered the

route of administration and

administration rates

Etc this is also the opportunity to

evaluate extreme values in the

observations when one does export data

analysis in particular

visualizations this these become

apparent one we also want to evaluate

the proportion of measurements that fall

below the limit of

quantitation a determination of samples

that are missing or damaged so these are

two separate components below the limit

of quantitation means a measure

measurement was made but the limit of

the quantitation of the assay was um

crossed and

therefore uh the measurement

value uh may be reported just as below

the limit of quantitation or a number

that does not conform to the stringency

of the assay as it was developed for

quantifying those

values again a preliminary evaluation of

observations with an evaluation of

outliers potentially evaluating the

concentration time profile if we have

sufficient sampling

a preliminary determination of the
potential of the structural models to
test and the preliminary calculation of
potential parameter values overall
clearance for example the volume
distribution to use the starting values
all population PK analyses are nonlinear
regression problems and you have to

start with initial

values initial

estimates so for this exploratory data
analysis as I mentioned we have the
copine concentration measurements in the
controlled antic psychotics trial of
intervention Effectiveness Katie this
was a treatment study with patients in a
schizophrenia arm and an Alzheimers
disease arm and sparse concentration
measurements were taken had approxim
just over two concentration samples per
individual so as we explore the data
some tabular output in terms of the
population

description so number of

observations in the schizophrenia arm we

concentration observations and the
Alzheimers are so 9 observations
all
told the bottom table shows a
distribution by selfreported
brace uh and uh so you can evaluate the
relative contributions here and whether
or
not they appear to be uh representative
of a population at
large uh one thing to keep in mind as we
go through these different elements
is whether or not there appear to be
systematic differences between the arms
of the study or even amongst the
covariants U this may potentially
confound identifying a specific effect
if there are strong
correlations similarly here we have a
table top showing the distribution uh BX
and the schizophrenia and the
Alzheimers arms we had
0 uh individuals and weve got in
the schizophrenia arm 9 in the
Alzheimers arm uh in the schizophrenia
arm it was predominantly male almost

of the of the individual patients the
Alzheimers uh study was more evenly
split with actually slight majority of
female
patients in terms of smokers and
nonsmokers see that uh we had a
majority of smokers in the schizophrenia
arm
uh versus the Alzheimers arm so this
could be an issue because it was an
effect of smoking you might have smoking
and and arm confounded because of the
different distributions so this is
something to keep in mind when youre
exploring
explanatory
variables in terms of the weight
distribution you can see as well the
schizophrenia arm the weights
were substantially
higher uh similar ly for the age
distribution in the bottom table we have
uh Alzheimers course arm the age was
substantially higher not surprising with
respect to the type of study so again if
youre interested in in if you detect an

age Effect one has to consider that
there may be a different disease process
as well these are different arms of the
study dosages were different this is
these are all elements to keep in
mind so while tabular approaches are
helpful it also is important to
visualize the

information so this next slide which was
created using R and a package called
ggall shows a matrix plot for patient
demographics this is for both the
Alzheimer and schizophrenia studies
combined so we you see on the diagonal
of this

Matrix are the uh distribution for
example for age
so you can see it looks quite
bodal uh the distribution for weight
its a bit of a shoulder here so maybe
not entirely glossi in you see also the
distribution by
sex female and
male uh and you see the distribution of
smokers in the combined studies you see
thats almost 00 you combine the

studies in the off diagonal elements
were able to
evaluate
correlations and other aspects of the
distribution so you see a
relationship with sex and age across the
studies smoking
status and age as
well uh and similarly await distribution
difference for uh sex with males being
SL the
heavier uh smokers and weight do not
show at least for the combin set this
element in terms of the other off
diagonal elements here the size of these
squares represents the number of
individuals here we have nonsmokers and
smokers by
sex so males and and
females okay and then in the bottom left
set of off diagonal elements we actually
see the individual histogram so you can
see the distribution of values for
example so for age BX okay for
weight right
BX okay you can see the shift here this

is these are the males these are females

males typically being a bit heavier but

you can get a sense of signal

ranges now one of the major components

is that we have these we have these

subcategories so what happens when we

subset these characteristics lets say

by

by by by sex and smoking thats whats

shown in this next set of

plots and again on the diagonal we have

the distributions but this is now by

different sex groups you can see that we

have more females in the older group but

in the younger

group the females and males are pretty

much

matched similarly for

weight the males are somewhat

heavier

okay

um and then

by uh smoking

status uh you see

that uh males tend to be more likely to

be

smokes similarly if we subset by smoking

by by smoking and

nonsmoking you can see that most of the

smokers are in the younger age group

weights overlap but we again see this

big difference by uh by sex so many more

male smokers for

example

similarly uh the you again have the

distributions for the individuals and

look at in terms of

correlatives

um we have there is a significant

difference here so if we

consider uh sex and age have nonsmokers

and smokers theres a similar uh Trend

here so we have uh typically younger uh

individuals smoking in this in this

study set um by weight again weight and

you you see that um for smokers and and

nonsmokers in in females as smok a

little bit uh uh light

than in males but it doesnt appear to

be much difference in weight so we start

to get some insights about the

distribution of these

parameters so what if we
consider the study arms
separately so actually going back to the
previous slide you can see that there
was some you know there were some
differences there were some sex and you
know weight differences in sex and
smoking status and that sort of things
so if we go and look at only the
schizophrenia
arm notice now that the age
distribution is quite
different its
less uh it still has a shoulder but its
less sort of
bodal and the uh sex and age
differences are less
pronounced smoking and age differences
are less pronounced individuals tended
to be younger we still have slightly
heavier males and females in this
study um and so that might be important
information is there an effect of the
disease or is it the effect of the age
group or is it effect of the weight if
you're looking at specific effects on

pharmacokinetic parameters at the
population
moment you subset
this by sex
groups or by smoking and nonsmoking
groups see that many of these
differences are not preserved looks
previous previous graph so again age
weight smoking and along this Matrix
here you see the these distributions are
largely overlapping at least in terms of
the continuous
variables uh when you subset along those
status within the
study similarly you can see
some uh weight uh difference sub by
smoking and nonsmoking groups nonsmokers
smokers um Nails actually tend to be a
bit
lighter what about the Alzheimers arm
so here again you see that the age now
has shifted to the
right the weight now again does not have
as much of a shoulder looks more
gossan the uh sex distribution is more
even and the smoking distribution is

quite

dramatically different very few smokers

actually in this uh in this

study and if we subset by the uh sex and

smoking

status you see that for the different uh

sex

groups um you can see that theres

almost

no uh smokers

here for female versus

male and

that we do have a shift in the weights

females are lighter than males in this

case ages though are quite closely cing

so things to keep in mind in terms of

potential confounders again this is done

with G galy and its I think an

important uh evaluation of the data

before we jump into the

[Music]

analysis so those are patient

demographics what about par what about

elements that are related to the pH

cinetics

directly so these tables show Doses and

the frequency by trial arm so the top table focuses on the dosage interval so there were hour dose intervals and hour dose intervals we can see that in the schizophrenia study the majority of individuals receive the doses twice a day of those individuals receive the doses every hours in the Alzheimers arm no patients receive doses every hours 00 of patients receive them every hours now this might be of concern if the drug has a relatively fast half-life perhaps you'll have more measurements for aluminum quantitation because of the dosage interval being wider before the patients are redosed overall if we combine the studies about 0 of the patients were receiving uh kipine every hours and were receiving them every hours some patient idea IDE some patients had both dose intervals

implemented so they had some periods
when they were on every 4 hours and
some periods when they were on every
12 hours what about the doses administered

so this uh bar chart is showing doses
daily doses for the schizophrenia arm
you'll see that we have doses that go
from

uh two uh 200 to 200 uh milligrams in
the schizophrenia on not very many low
doses so this would be 200 milligrams
twice a day for example these were all
given every 4 hours um and a relatively
even dist

distribution uh across the the dosage
ranges so well represented for the
analysis with respect to the Alzheimers

disease arm you see the dose
distribution is quite different its
well given every uh 12 hours and the
doses R dosages range from 25 to 200
milligrams daily the doses are listed
here in micrograms that matches our
units in the actual analysis and the
majority of doses here at the low end
milligrams so are we going to see many

more again limited quantitation issues
with the Alzheimers disease on
because theyre lower doses over a wider
interval then again theyre older
patients and maybe they eliminate the
doses more slowly so these are all
things to keep in mind with respect to
uh the population pharmacokinetic

[Music]

analysis another key element to evaluate
is how much information do we have what
is how many concentration measurements
do we have across the entire
concentration time profile we have
absorption processes distribution
processes in elimination processes and
these are sparse samples so its
important to understand whether or not
you have sufficient sampling across the
entire dosage interval so that you can
inform those parameters so on the left
this is from the schizophrenia arm we
have the time since last dose
distribution with a hour dose
interval we have a frequency on the y
axis and the time since last dose on the

xaxis what you see is that we have quite a few samples shortly after the do so well get the absorption processes reasonably well but as we get closer to the end of the interval we have fewer samples less information with respect to the time since last dose distribution for the hour dosage interval you can see we again have some we have less information actually in the absorption profile and then we have more information in the tail phase of the distribution V concentration time profiles for the Alzheimers disease study we see a similarly bodal distribution which probably relates to clinics being closed in the middle of the night and patients coming in for usual clinic hours and depending on whether or not they take their their dose in the morning or the evening again time since last dose we have the frequency on the y axis time since last dose on the x axis you see that we have information that will that

will help understand the absorption
process absorption profile
and then the tail distribution as we uh
uh go towards the end of the dosage
interval or reasonably well uh reasonably
well
captured so what about actual
concentration measurements well heres a
plot with the uh dose normalized
concentration remember we had multiple
concentrations this is for the
schizophrenia
arm and each of these black lines
represents uh
an individual with con with the time
after time since last dose on the xaxis
here time since last dose and theyre
connected so you can see profiles for
every
individual the green line in the Shaded
area is basically a
smooth
through these profiles so you can see
that
overall we capture a the up slope and
down slope of the concentration time

profile that one would expect with this
uh with this drug and we can evaluate
whether or not there appears to be
multiexponential character we can also
evaluate whether or not this may be
sufficient to actually allow us to
calculate some preliminary initial
estimates for clearance and volume
distribution for

example now there were multiple doses
given though and maybe there's a
nonlinearity in dose to evaluate that
again we have dose normalized
concentration on the y axis and time
since last dose on the x axis but now we
have a color coding for the different
doses again for within individuals we
have connectors connecting the
doses and blue the light blue are the
highest doses the black are the lowest
Doses and you can see that there's pretty
much overlap so if you normalize the
concentrations it suggests that linear
processes are driving the
pharmacokinetic
profile

another evaluation is to look at the individual patients time concentration profile so this Matrix of plots is showing concentration on the y axis times the last do on the x axis again this is a schizophrenia study for for the hour interval and this is useful for identifying whether or not there appear to be any significant outliers uh by individual and we dont really see any here similarly we look at the hour interval you see this is for the schizophrenia arm and again dont necessarily see any outliers there a few maybe maybe higher concentrations here that we want to investigate uh but generally speaking uh reasonable for the Alzheimers disease arm here a sample of patients again this is a matrix plot why each panel represents a different individual y axis is the dose normalized concentration and the xaxis is the time since last dose and again you can use this to inspect for potential

outliers so in conclusion a complete
exploratory data analysis is critical
prior to beginning the actual model
building process for a population from
kinetic
analysis insights on what types of model
structures can be supported what
covariants may be considered and what
potential confounds can be
preidentified
and so our now that weve done the
exploration of the
data we can begin the population
pharmacokinetic analysis and this will
be covered in part in the vignette part
two Id also like to
acknowledge uh two individuals who
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much