

Prediction of within-subject variability using population approaches and its application to demonstrate highly variable drug

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PAGK Trainee session

I. Background and Objective

II. Method

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IV. Summary and Conclusion

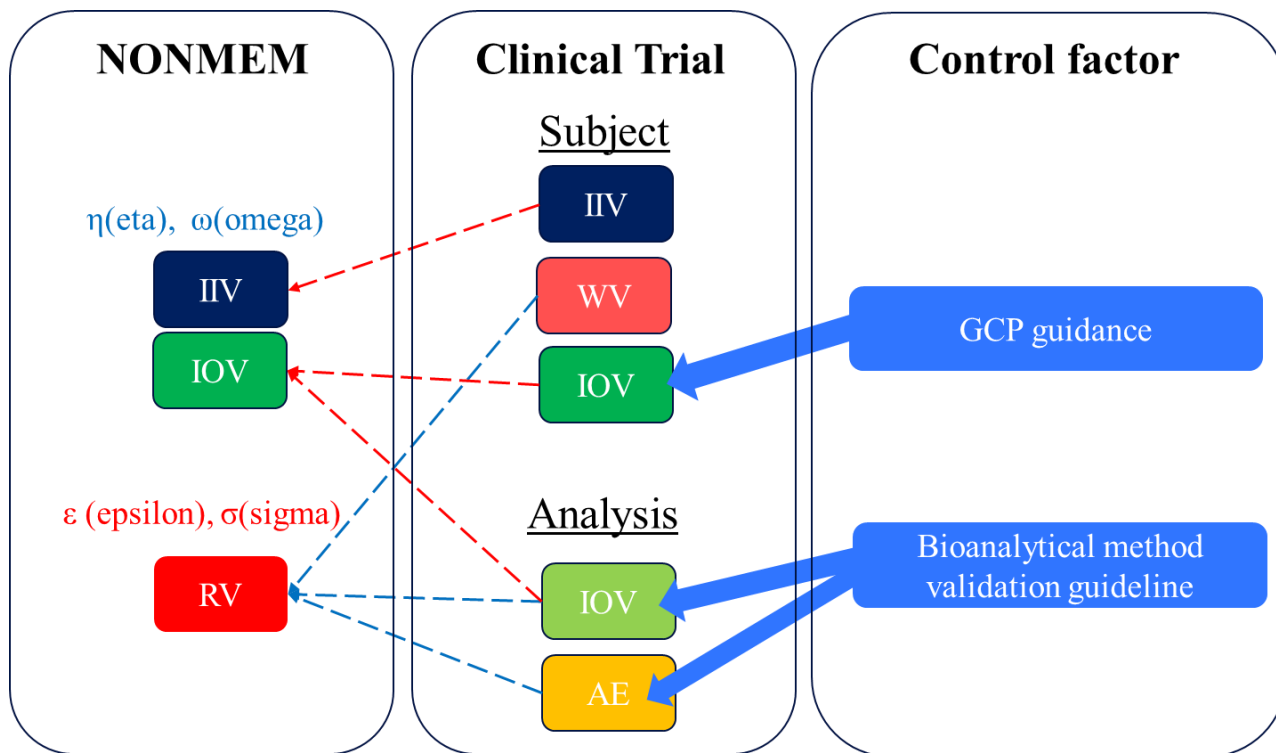
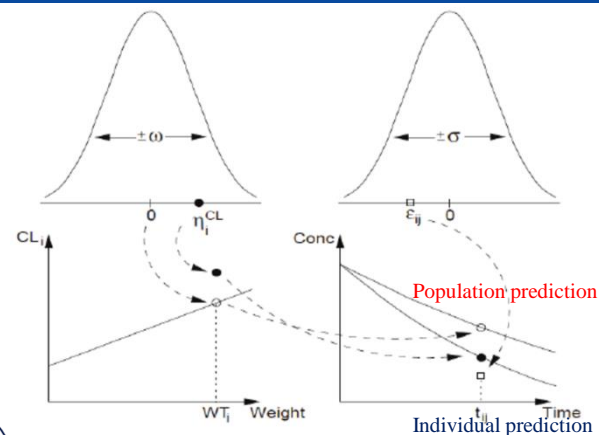
V. Appendix

I. Background and Objective

Random effect in Pop PK data

- **Unexplained differences between individuals**

- **Inter-individual variability, Between-subject variability, eta (η)**
- **Intra-individual variability, Residual variability, epsilon (ε)**
- **Inter-occasion variability**

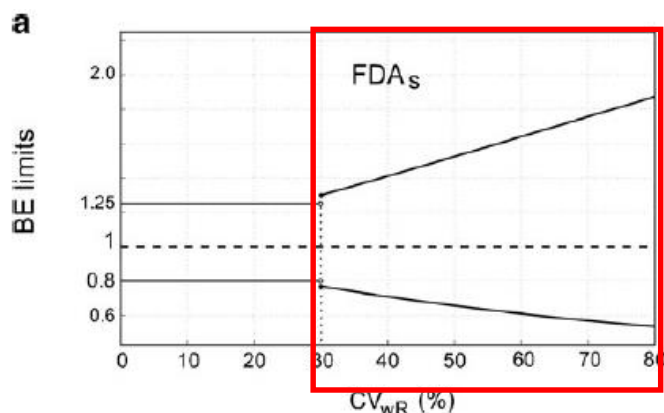


Background for highly variable drug

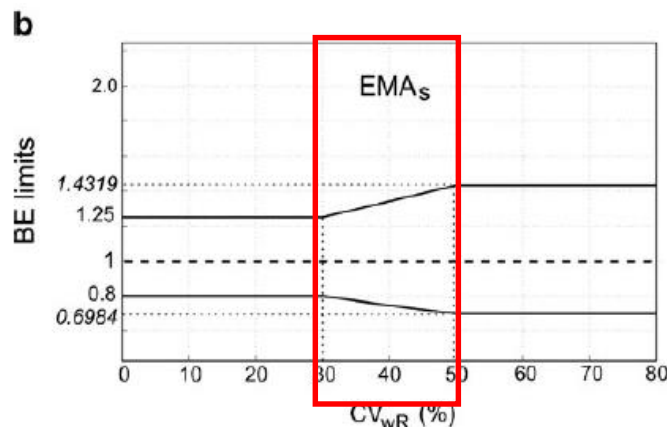
Definition of highly variable drug(HVD)

HVDs : drug products exhibiting within-subject variability of 30% (CV_w , coefficient of variation) or greater in the pharmacokinetic measures AUC and/or C_{max}

Widening of BE limited based on reference variability _ FDA vs EMA



$$Upper/Lower limits = \exp \left(\pm \ln(1.25) \cdot \frac{s_{wR}}{s_{w0}} \right)$$



$$Upper/Lower BE limits = \exp (\pm k \cdot s_{wR})$$

- 1. Verification how well NONMEM can estimated residual variability through simulated population pharmacokinetic dataset under various condition**
- 2. Confirmation that this population approach can be applied to the real highly variable drug case.**

II. Method

Overall experiment scheme

A. Experiment 1 (5 different levels of WV(10%, 20%, 30%, 40%, and 50%) without IIV's change(0%))

Generating of simulation data sets using by R

1000 simulation data sets

- CL=10 L/hr, V=50 L
- AMT = 100 mg
- 12 sampling point
= 0, 0.083, 0.167, 0.333, 0.5, 1, 2, 4, 6, 8, 12 & 24 hr
- Subject No. = 6, 12, 18, 24 & 30
- WV = 10, 20, 30, 40 & 50%
- IIV = 0%

PK Modeling execution using by NONMEM

I.V. PK model

- 1 compartment I.V. modeling
- Proportional error model
- FOCE with interaction estimation

Comparison RV with WV

Comparison two values

- WV established in R code
- RV estimated by NONMEM
- Success data for approximation of RV to WV

B. Experiment 2 (5 different levels of WV(10%, 20%, 30%, 40%, and 50%) with IIV's change(10→50%)

Generating of simulation data sets using by R

1000 simulation data sets

- CL=10 L/hr, V=50 L
- AMT = 100 mg
- 12 sampling point
= 0, 0.083, 0.167, 0.333, 0.5, 1, 2, 4, 6, 8, 12 & 24 hr
- Subject No. = 6, 12, 18, 24 & 30
- WV = 10, 20, 30, 40 & 50%
- IIV = 10, 20, 30, 40 & 50%

PK Modeling execution using by NONMEM

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- 1 compartment I.V. modeling
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- WV established in R code
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- Success rate for approximation of RV to WV

Application for real case(e.g. eperisone)

Pharmacokinetic and bioequivalence study of sugar-coated and film-coated eperisone tablets in healthy subjects, Hyun-Ju Lee et al., International Journal of clinical pharmacology and therapeutics, Vol. 57, No1(55-62), 2019

R scaled approach

G/P	P1	P2	P3
A (N=12)	R	R	T
B (N=10)	R	T	R
C (N=11)	T	R	R

R : Murex[®] 50 mg, Cho Dang Pharm Co., Ltd

T : Eperex[®] 50 mg, Korea United Pharmaceutical Co., Ltd

Result : Geometric mean ratio, 90% confidential intervals and within subject variability for AUC_t and C_{max} using the EMA method

Parameters	GMR	90% CI	SwR	CVwR(%)	BE limit
AUC _t	0.9836	0.8275 ~ 1.1692	0.332	33.17	0.8 ~ 1.25
C _{max}	0.9402	0.7587 ~ 1.1652	0.474	50.21	0.6984 ~ 1.4319

Work flow

Random sampling of PK dataset
(N =6, 12, 18, 24, 30)



Estimation PK parameter
& Sigma(σ) value



Visual prediction check

**Random sampling from reference drug's
PK data**

**PK modeling : 1 compartment, oral absorption,
first-order elimination**

Model diagnostic

III. Results

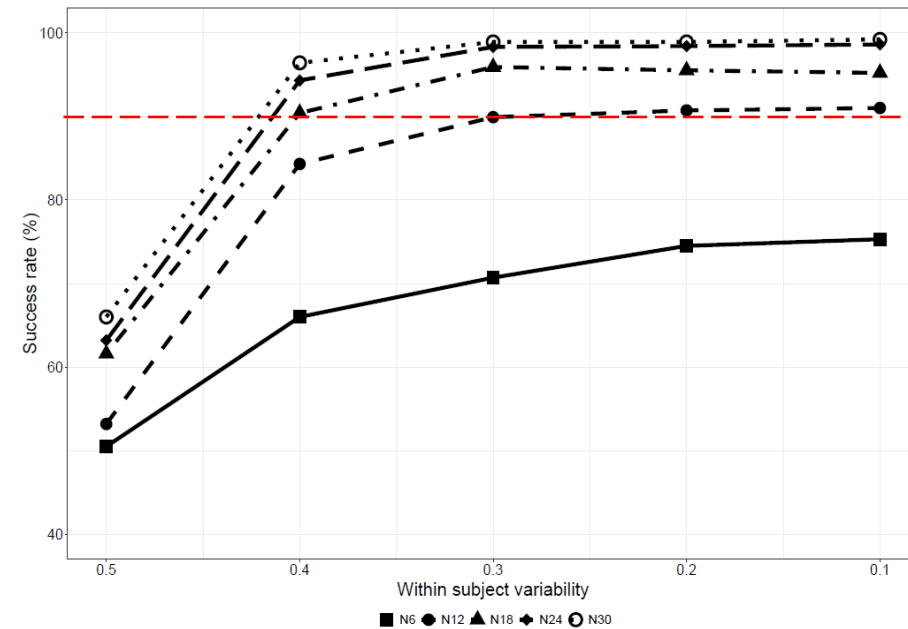
Result for Experiment 1

(5 different levels of WV(10%, 20%, 30%, 40%, and 50%) without IIV's change(0%)

Setting WV(%)	Success rate(%)* for each subject number				
	N=6	N=12	N=18	N=24	N=30
10	75	91	95	99	99
20	75	91	96	98	99
30	71	90	96	98	99
40	66	84	90	94	96
50	51	53	62	63	66

Tabulated summary for results of first experiment

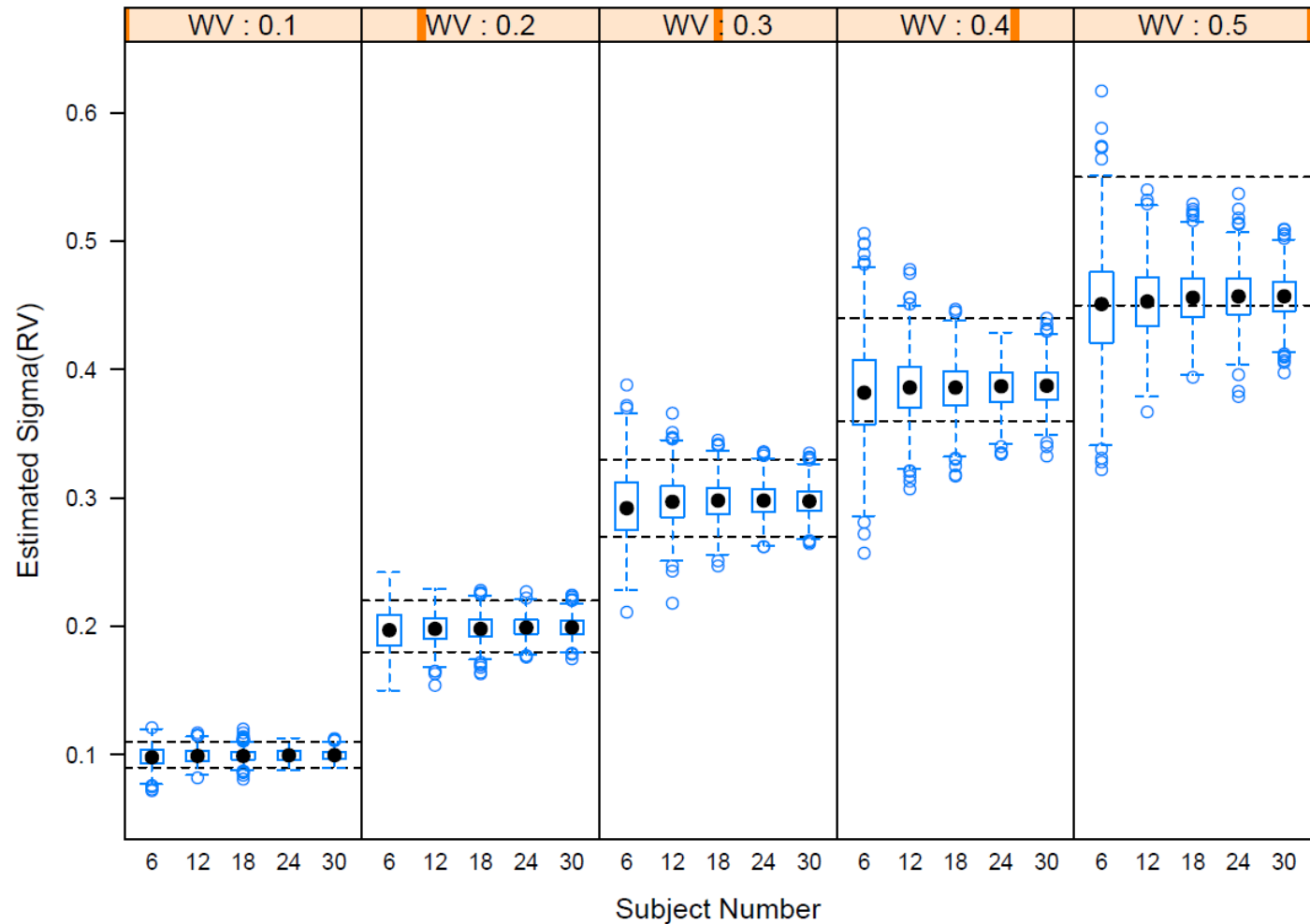
*Success rate at which estimated sigma values are included in True value(Setting WV values) $\pm 10\%$



Result for Experiment 1 _ Cont'd

(5 different levels of WV(10%, 20%, 30%, 40%, and 50%) without IIV's change(0%))

Within Subject Variability



Result for Experiment 2

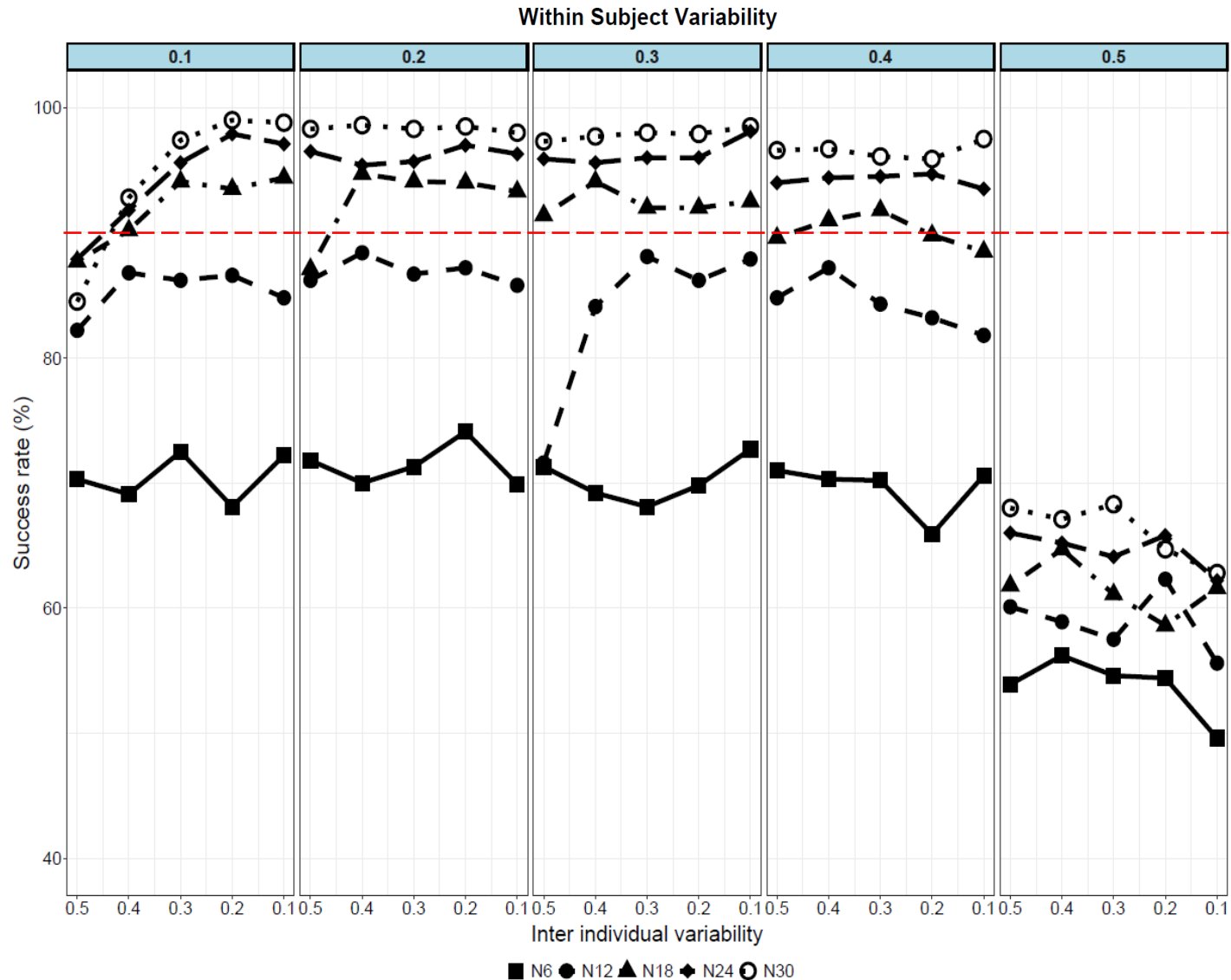
(5 different levels of WV(10%, 20%, 30%, 40%, and 50%) with IIV's change(10→50%))

Setting Condition		Success rate(%)* for each subject number				
WV(%)	IIV(%)	N=6	N=12	N=18	N=24	N=30
10	10	72	85	94	97	99
	20	68	87	94	98	99
	30	73	86	94	96	97
	40	69	87	90	92	93
	50	70	82	88	88	85
20	10	70	86	93	96	98
	20	74	87	94	97	99
	30	71	87	94	96	98
	40	70	88	95	95	99
	50	72	86	87	97	98
30	10	73	88	93	98	99
	20	70	86	92	96	98
	30	68	88	92	96	98
	40	69	84	94	96	98
	50	71	72	91	96	97
40	10	71	82	89	94	98
	20	66	83	90	95	96
	30	70	84	92	94	96
	40	70	87	91	94	97
	50	71	85	90	94	97
50	10	50	56	62	62	63
	20	54	62	59	66	65
	30	55	58	61	64	68
	40	56	59	65	65	67
	50	54	60	62	66	68

*Success rate at which to estimated sigma values are included in True value(Setting WV values) \pm 10%

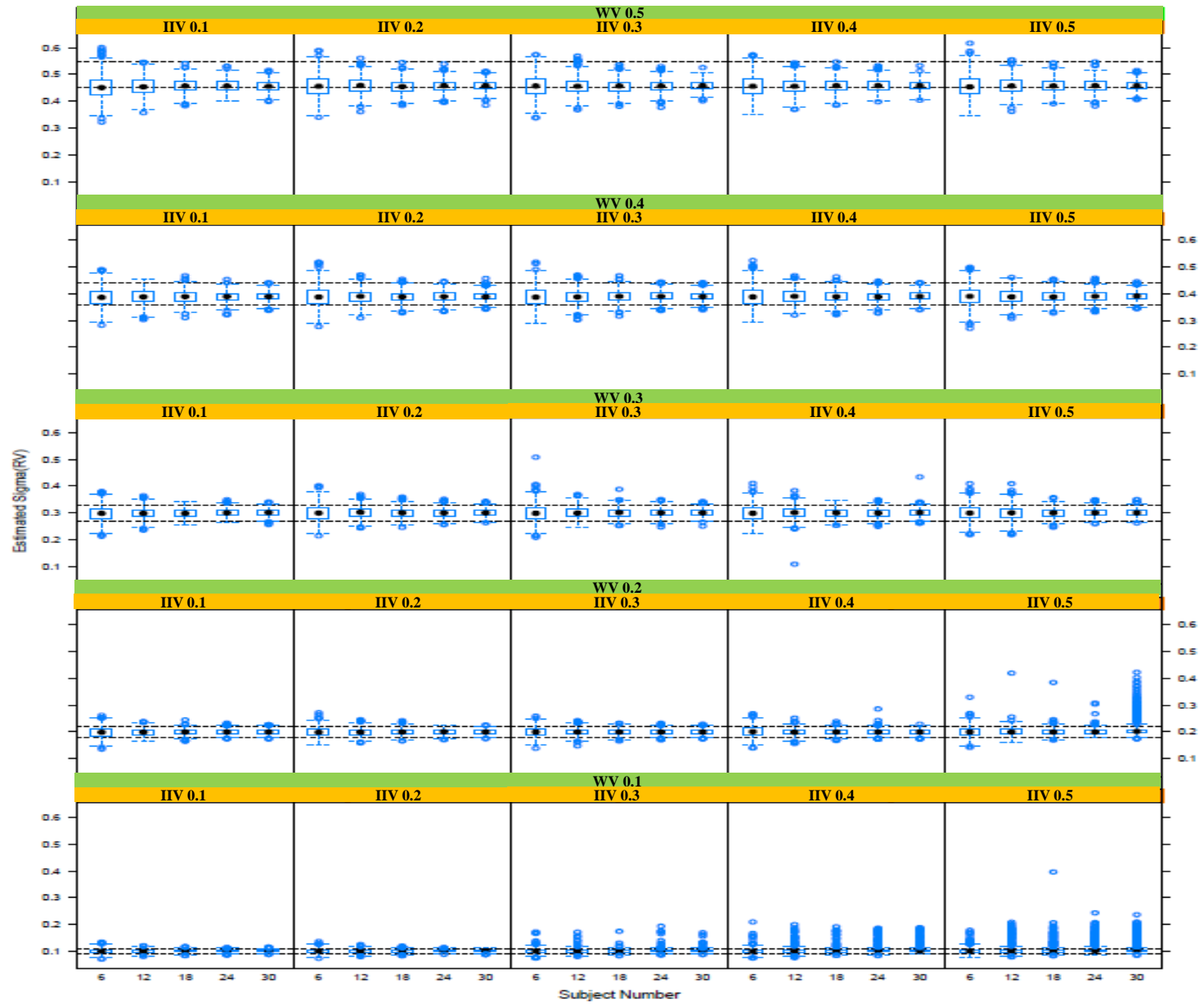
Result for Experiment 2 _ Cont'd

(5 different levels of WV(10%, 20%, 30%, 40%, and 50%) with IIV's change(10→50%)



Result for Experiment 2 _ Cont'd

(5 different levels of WV(10%, 20%, 30%, 40%, and 50%) with IIV's change(10→50%)



Result for real case application

The result for real application

Subject No.	6	12	18	24	30
Sigma, σ (%)	44.9	47.7	44.5	43.8	47.2

Cf. Result from original reference

Parameters	GMR	90% CI	SwR	CVwR(%)	BE limit
AUC _t	0.9836	0.8275 ~ 1.1692	0.332	33.17	0.8 ~ 1.25
C _{max}	0.9402	0.7587 ~ 1.1652	0.474	50.21	0.6984 ~ 1.4319

Pharmacokinetic and bioequivalence study of sugar-coated and film-coated eperisone tablets in healthy subjects, Hyun-Ju Lee et al., International Journal of clinical pharmacology and therapeutics, Vol. 57, No1(55-62), 2019

IV. Summary and Conclusion

Summary and Conclusion

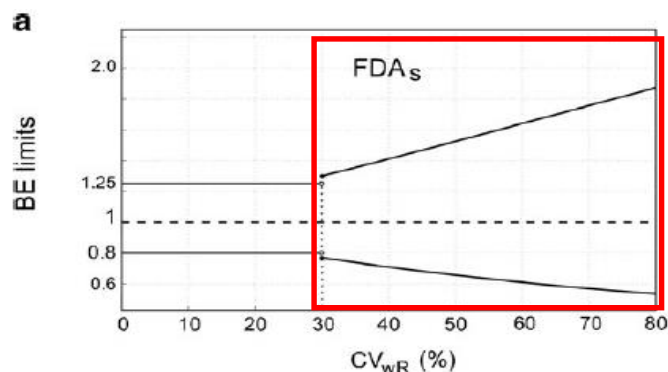
- **When the IIV was no change(0%)**
 - **WV 10~30%** : 90% or more prediction success rate with **12 or more** subjects
 - **WV 40%** : 90% or more prediction success rate with **18 or more** subjects
 - **WV 50%** : **Underestimation** at 6~30 subjects
- **When the IIV was change(10~50%)**
 - **WV 10~40%** : 90% or more prediction success rate with **18 or more** subjects
 - **WV 50%** : **Underestimation** at 6~30 subjects
- **Real HVD case(eperisone)**
 - **Our Pop. approach** result : **44 ~47%** for RV at which **6~30 subject** number
cf. BE result : 33.17% as a CV_{wR} for AUC and 50.21% as CV_{wR} for Cmax

In conclusion, we have confirmed that our methodology is relatively accurate in well-estimating within subject variability from population PK data. Also, we have confirmed that it can be used as a tool to judge the highly variable drug.

Thank you

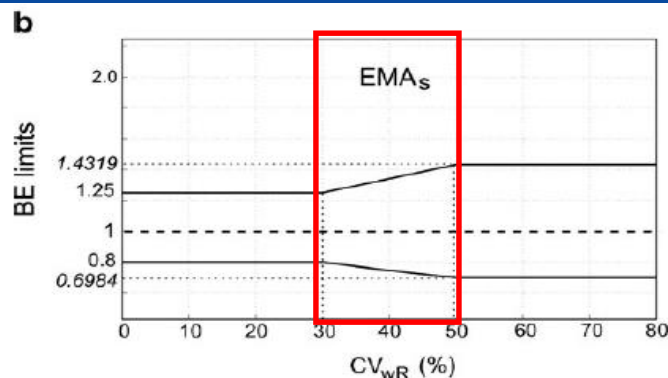
V. Appendix

Background for highly variable drug _ Cont'd



$$Upper/Lower limits = \exp \left(\pm \ln(1.25) \cdot \frac{s_{wR}}{s_{w0}} \right)$$

$s_{wR}(\%)$	BE limit	
	Upper	Lower
30	1.3070	0.7651
40	1.4291	0.6998
50	1.5625	0.6400
60	1.7084	0.5854
70	1.8679	0.5354
80	2.0423	0.4897



$$Upper/Lower BE limits = \exp (\pm k \cdot s_{wR})$$

s_{wR} : within-subject variability
(corresponding CV_{wR} value)

s_{w0} : switching variance, mainly 0.25

K : scaling factor, 0.760

Random effect in Pop PK data

_How to engage in generating simulation data

R script used in this study

```
##Individual PK parameters(IIV 10%)
data.sample <- 1:1000
start.number <- min(data.sample)
i <- start.number

for (i in data.sample) {
  set.seed(seed[i])
  x <- rlnorm(100000, meanlog = 10, sdlog = 1)
  y <- rlnorm(100000, meanlog = 50, sdlog = 5)

  CL <- sample(log(x)[log(x)>0],12)
  Vd <- sample(log(y)[log(y)>0],12)

  id <-seq(1:12)

  data <- data.frame(id,CL,Vd)
  write.table(data,paste("data",i,".csv",sep=""),sep=" ",row.names = F)
}
```

Give 10% as standard deviation to the mean values for x and y, respectively

Random effect in Pop PK data

_How to engage in generating simulation data_Cont'd

R script used in this study

```
##Calculation the concentration using parameters(CL, Vd)
total <- 1:1000
min.total <- min(total)
k <- min.total

for(k in total){
  one <- read.table(paste("data",k,".csv",sep=""),sep = ",",header=TRUE)
  kel <- one$CL/one$Vd
  one <- cbind(one,kel)

  number <- 1:12
  begin <- min(number)
  i <- begin

  for(i in number){
    special.ID <- DO[DO$ID==i,]

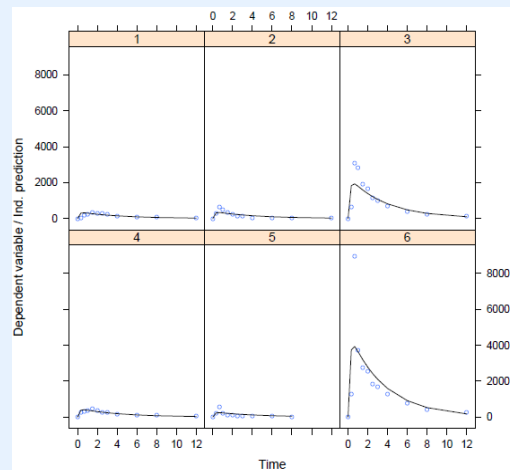
    timenumber <- 1:12
    time.start <- min(timenumber)
    j <- time.start

    for(j in timenumber){
      #iv PK equation
      conco <- (100*exp(-one$kel[i]*special.ID$TIME[j])/(one$Vd[i]))
      eps <- rnorm(12, mean = 0, sd = 0.3)
      eps.sample <- sample(eps,1)
      conco2 <- conco*(1+eps.sample) ##Proportional model
```

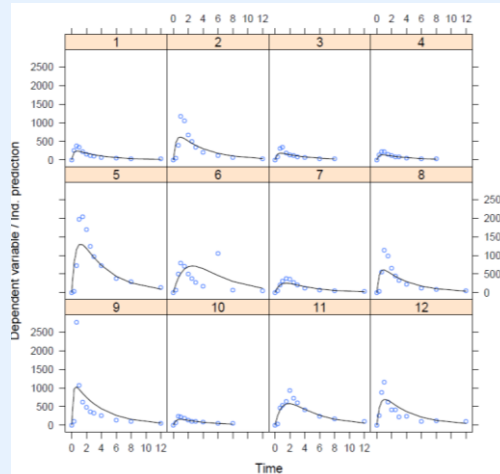
Give the calculated plasma concentration epsilon with a mean of 0 and a standard deviation of 0.3.

DV + IPRED vs Time _ Individual

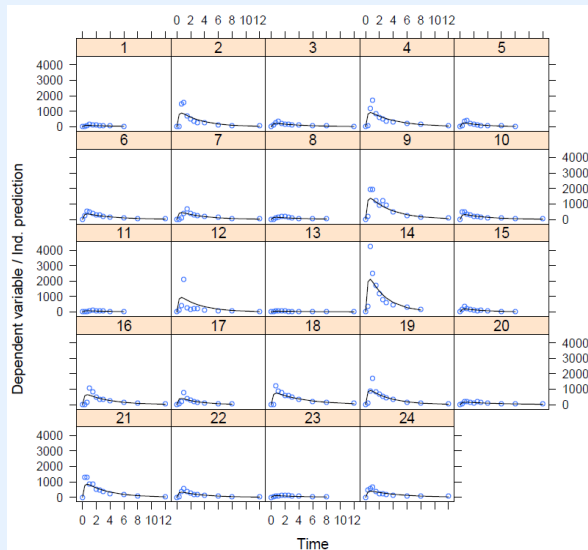
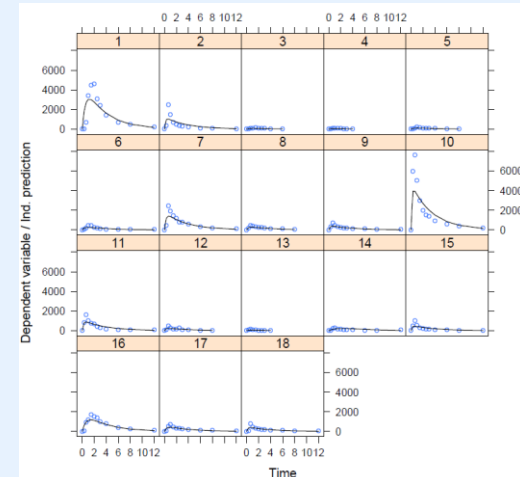
6 subject No.



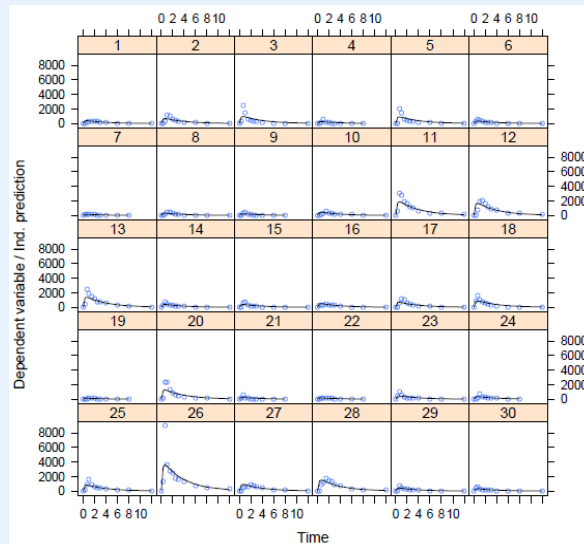
12 subject No.



18 subject No.



24 subject No.



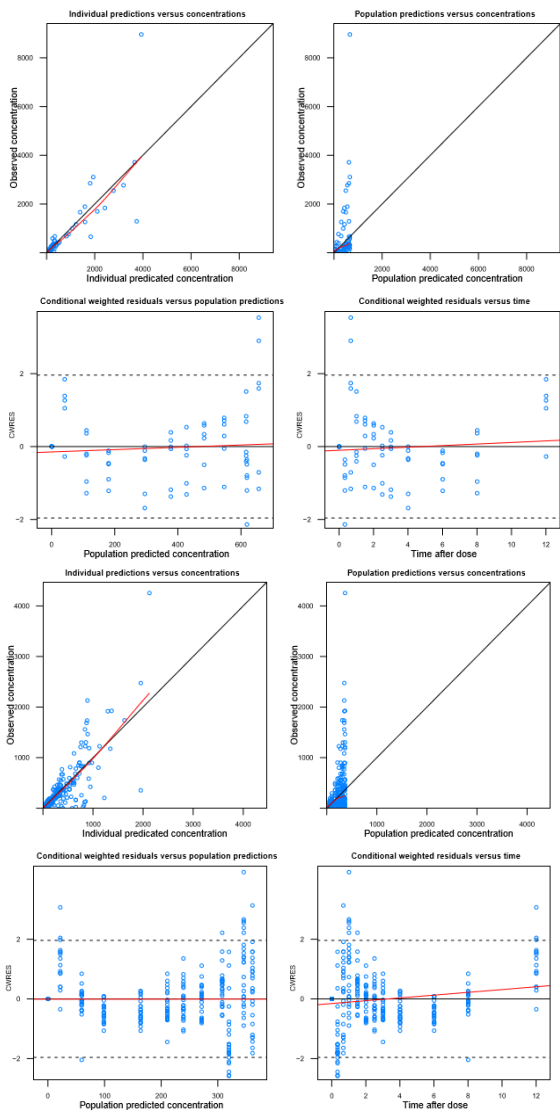
30 subject No.

DV
IPRED

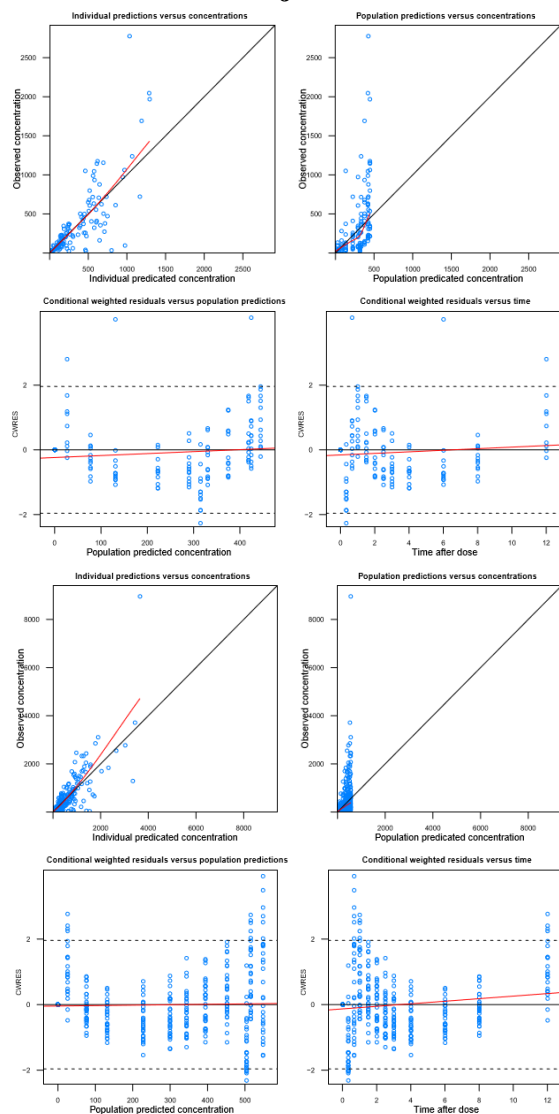


Result for real case application _ Basic goodness of fit

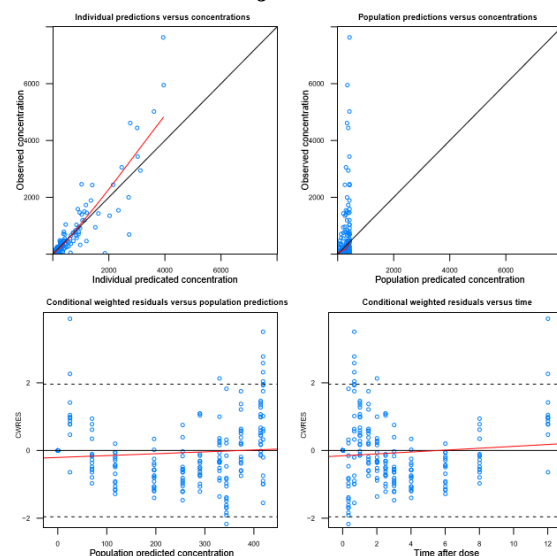
6 subject No.



12 subject No.



18 subject No.



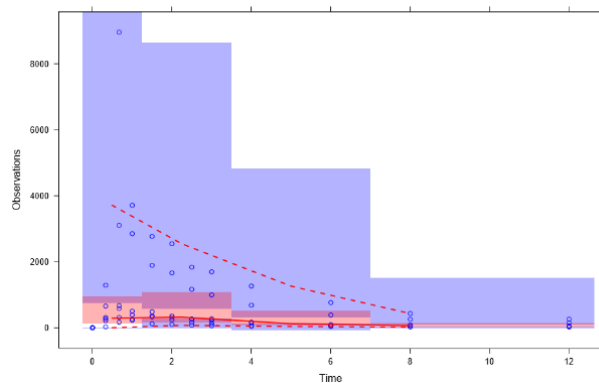
24 subject No.

30 subject No.

Result for real case application _ Visual prediction check

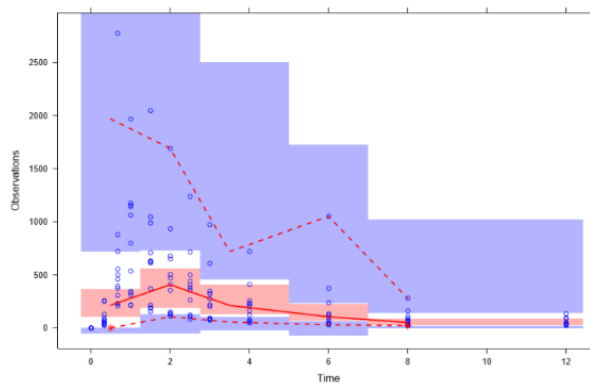
6 subject No.

Visual Predictive Check
Observations vs. Time (Run 0)



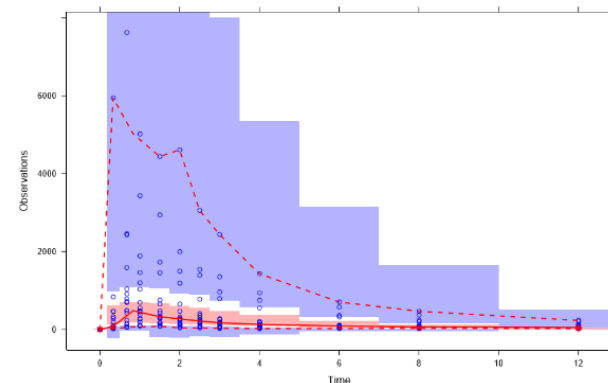
12 subject No.

Visual Predictive Check
Observations vs. Time (Run 0)



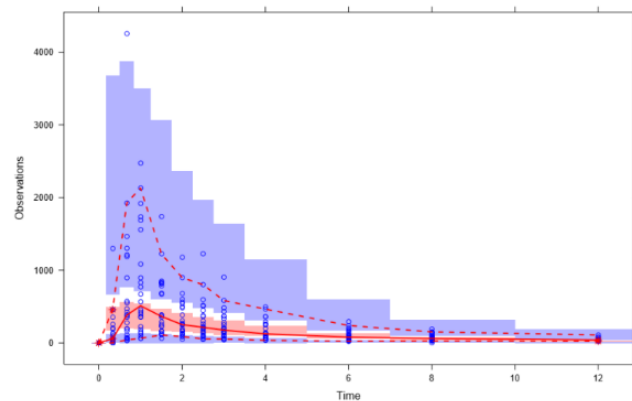
18 subject No.

Visual Predictive Check
Observations vs. Time (Run 0)



24 subject No.

Visual Predictive Check
Observations vs. Time (Run 0)



30 subject No.

Visual Predictive Check
Observations vs. Time (Run 0)

