

# The effects of an HIV partner on a patient's HIV severity

*Fan Yang, Jia Liu, Josh Wiltsie, Monica Patrin,  
Rodrigo Lovaton, Seongkwon Lyu*

# Introduction

- Problem: A researcher wants to know if partner's HIV status has an effect on three blood measurements for patients with HIV?
- Data includes: patient's gender, partner status (HIV or not), and three blood measurements (CD4, CD8, and RNA).
- $N = 278$  patients with HIV (83 missing values for RNA, 2 missing for CD4 and CD8).

# Responses

- **CD4** or white blood cell count - measures the severity of sickness. Higher is healthier.
- **CD8** cell count - measures the patient's ability to fight infections. Higher is healthier.
- **RNA** test - measures how much virus is in the patient's blood. Lower is healthier.

# Data collection

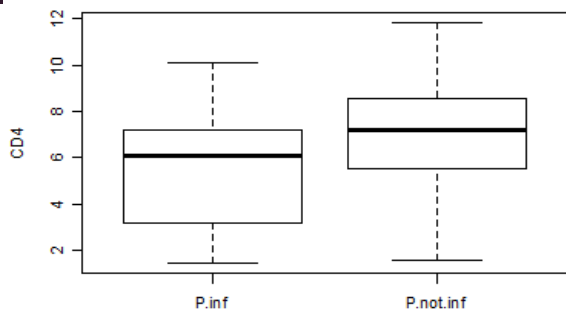
- The data was collected over a one-month period at a single clinic.
- Participation was voluntary. A gift card was given to all participants.
- The sample is expected to be similar to the overall HIV infected population of interest for the clinic.

# Background

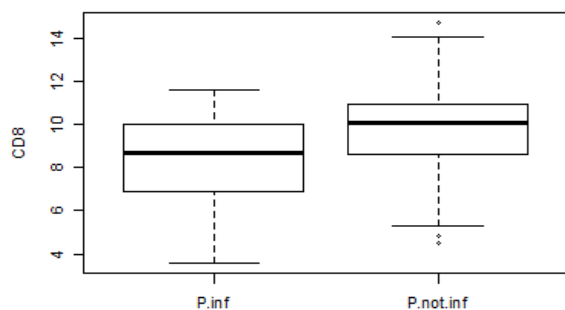
- Two other unmeasured factors could possibly affect the blood measurements:
  - Type of treatment being administered and compliance with treatment
  - Length of time since diagnosis
- Sampling bias is possible but the client thinks the sample represents the population well.
- Other demographics are not thought to affect responses (eg. ethnicity, income).

# Box plots for CD4, CD8, and log(RNA) by partner's HIV status and gender

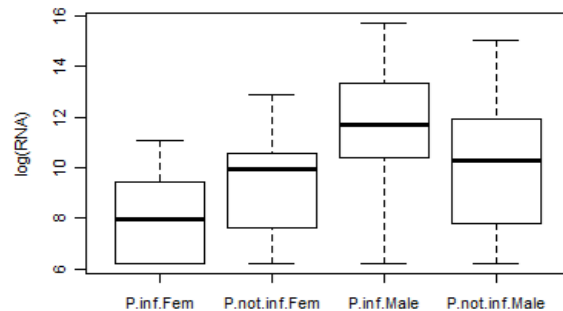
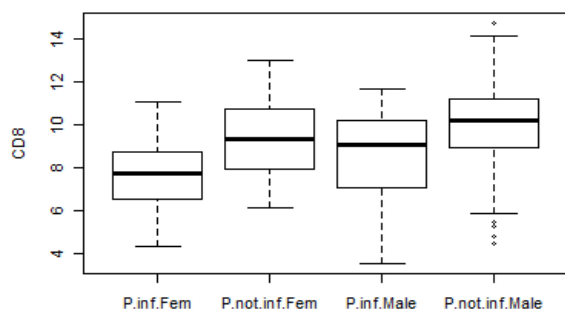
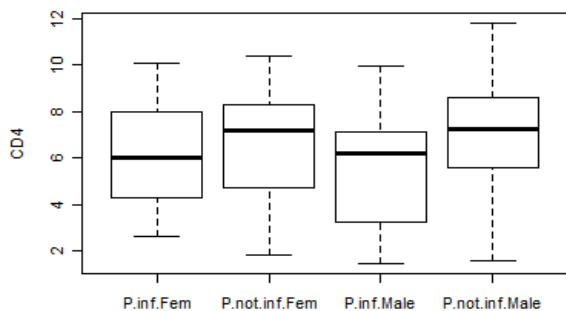
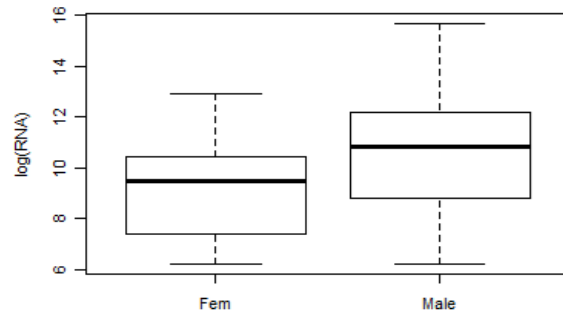
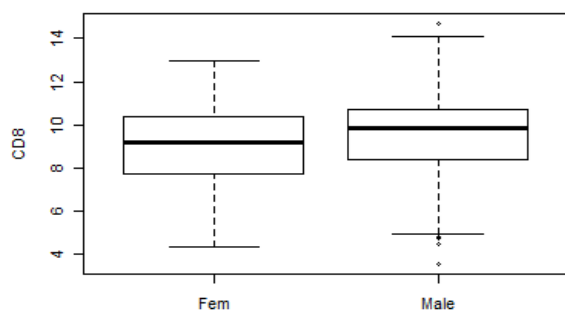
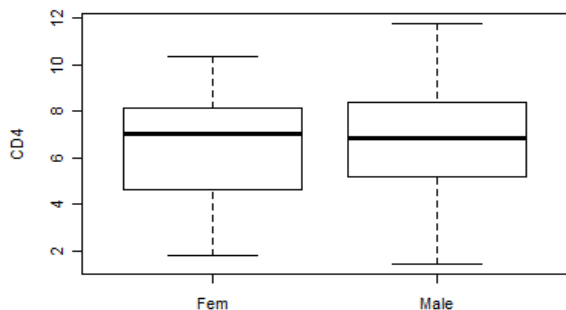
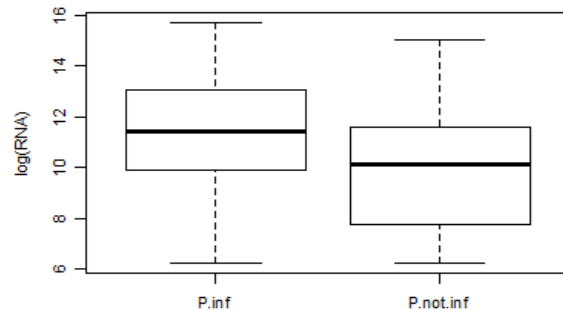
## CD4



## CD8



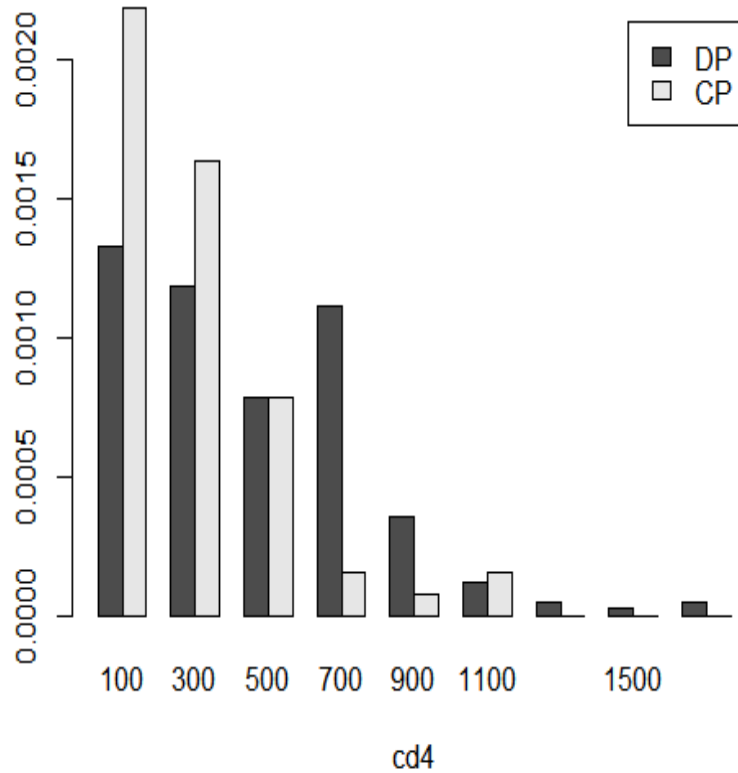
## Log(RNA)



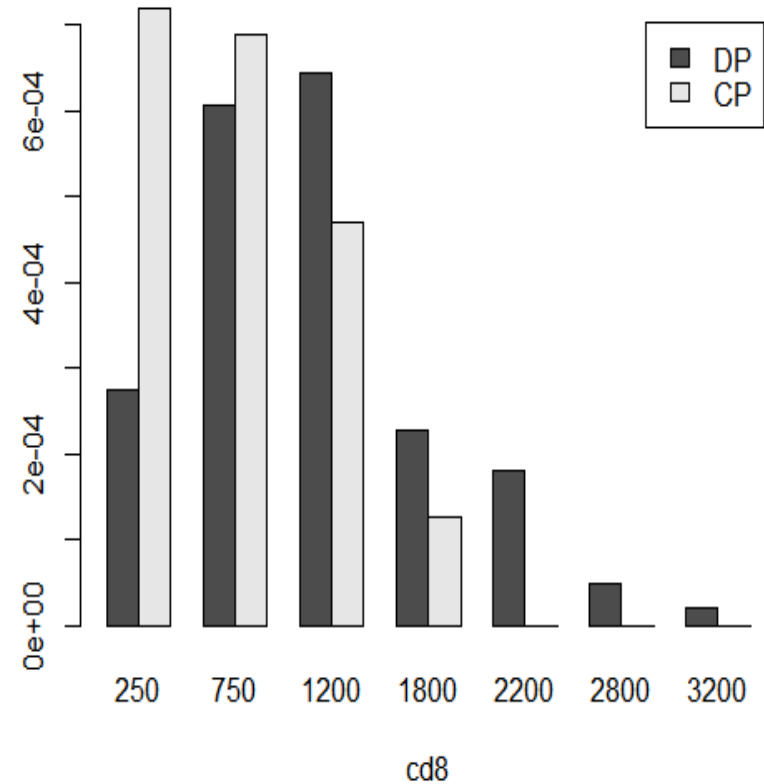
# Part II: Data Exploration

# Side-by-side histogram

Side-by-side cd4 histogram for two types



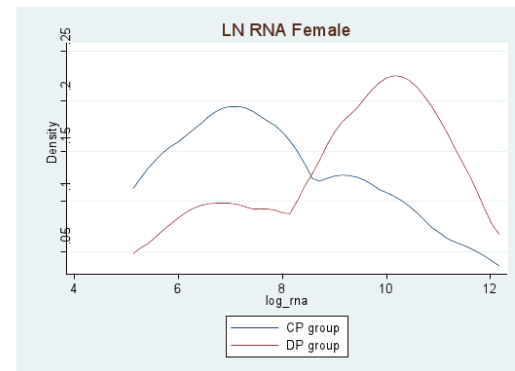
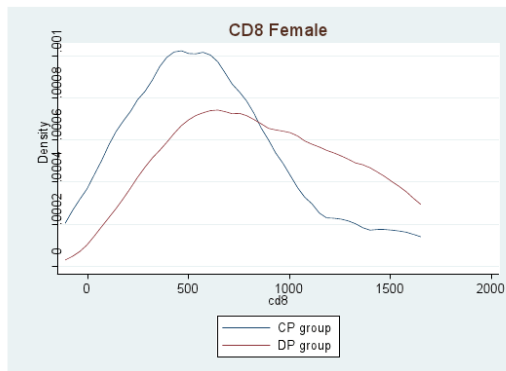
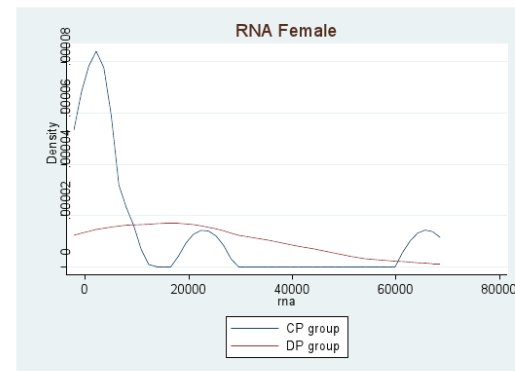
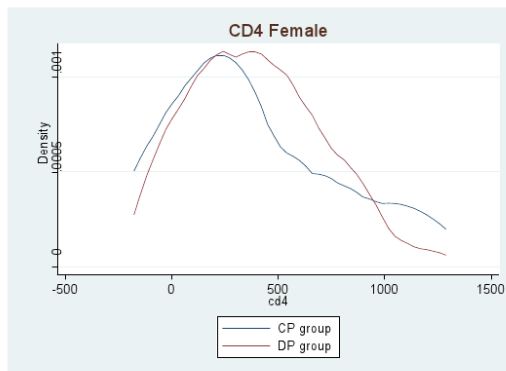
Side-by-side cd8 histogram for two types

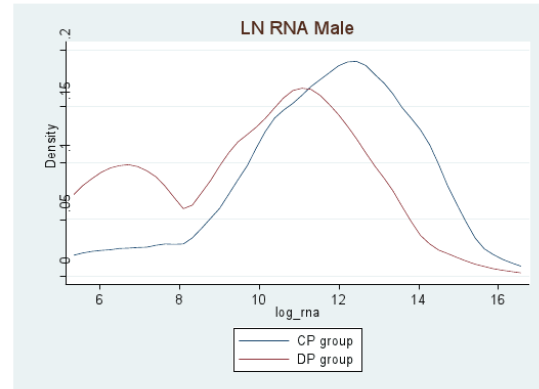
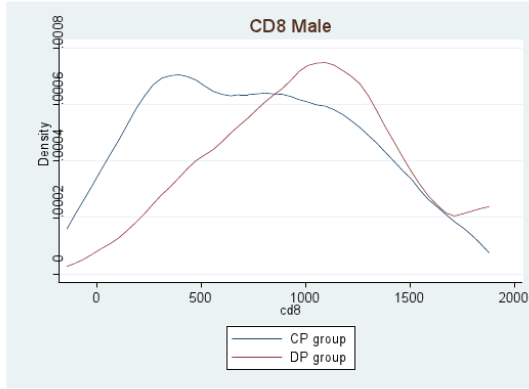
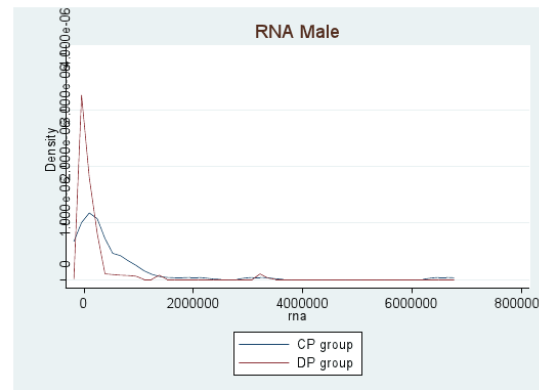
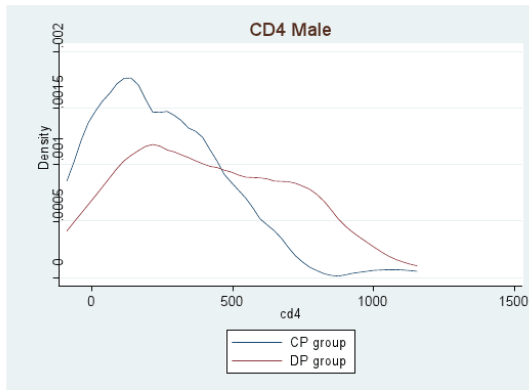




# Data visualization

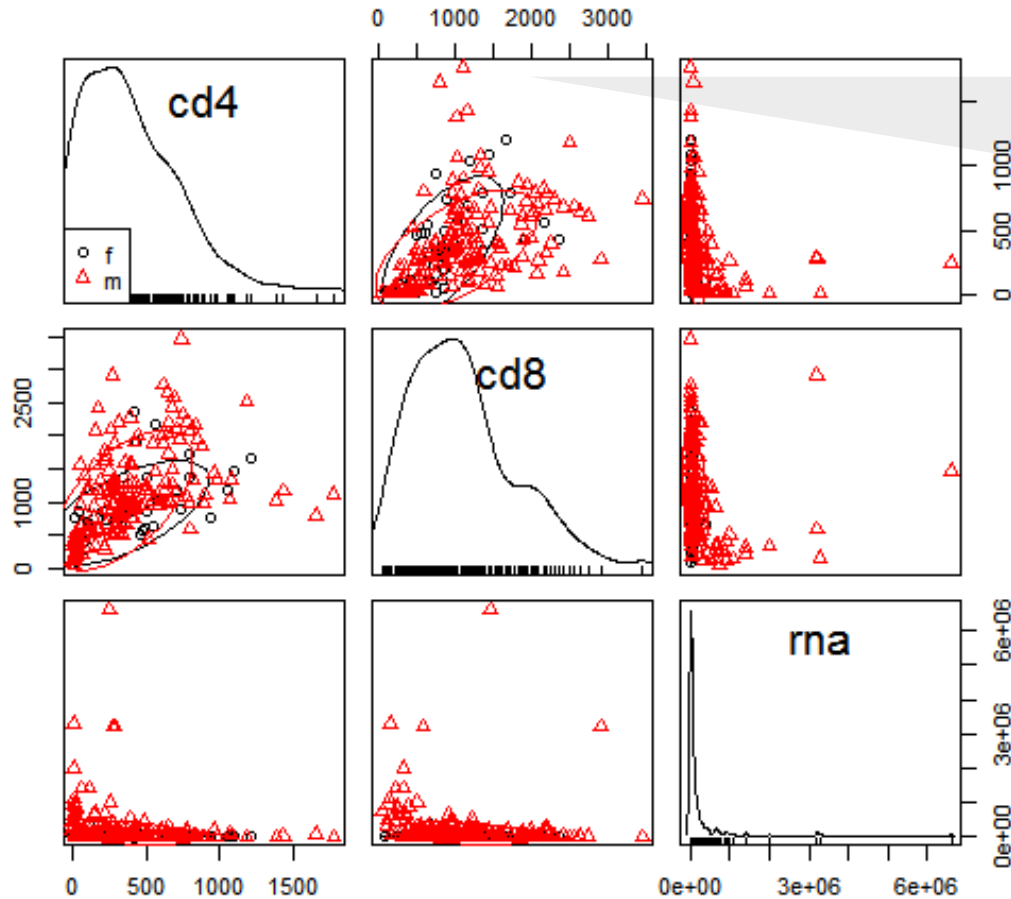
- Kernel density curves





From the graphs we can see that sex and type seem to have effect on some responses.

# Scatterplot matrix



RNA is very skewed,  
and the multivariate normality seems violated.

# Transformation

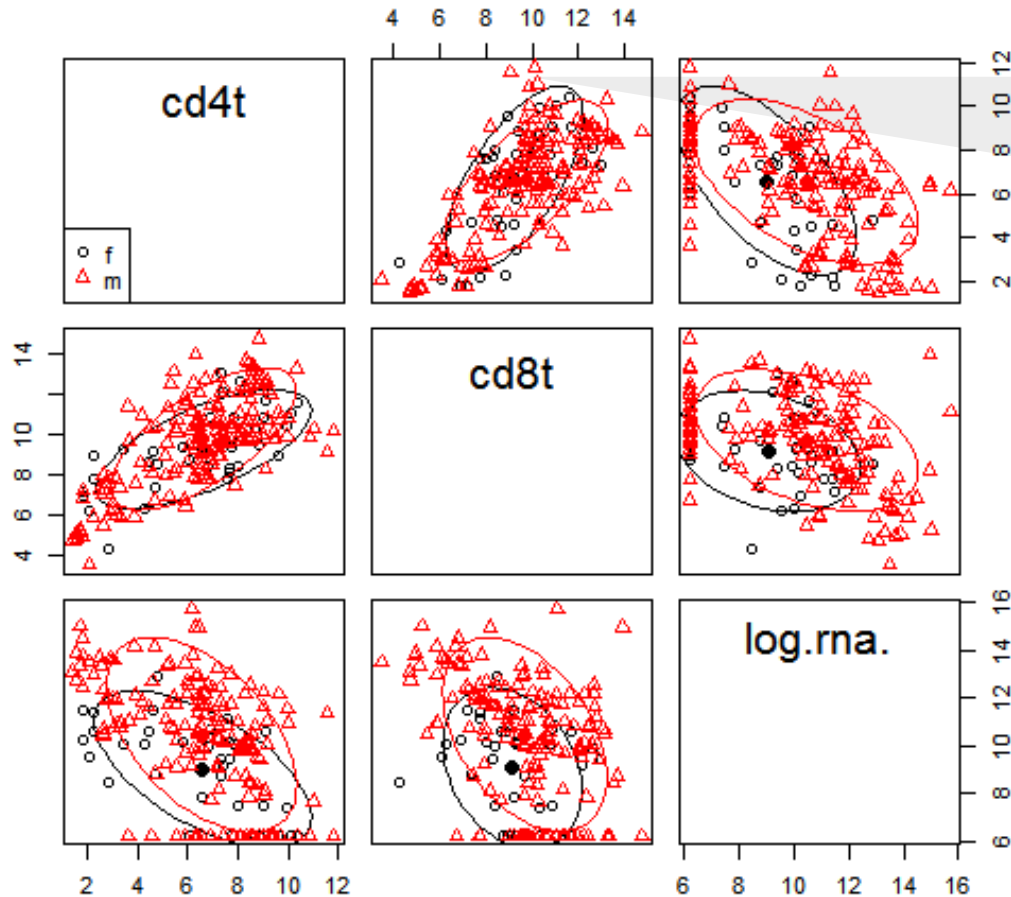
- For later analysis, we want our responses to be multivariate normal. Using R we figured out to do the following

```
> cd4 <- cd4^0.33
```

```
> cd8 <- cd8^0.33
```

```
> rna <- log(rna)
```

# Scatterplot Matrix



The multivariate normality assumption seem to hold now.

# Part III: Data Analysis

# MANOVA

```
> man1 <- manova(cbind(cd4.tr, cd8.tr, rna.tr) ~ type*sex, AIDS) )  
> summary(man1)
```

	Df	Pillai	approx F	num Df	den Df	Pr(>F)
type	1	0.136803	9.8260	3	186	<b>4.771e-06 ***</b>
sex	1	0.077059	5.1765	3	186	<b>0.001856 **</b>
type:sex	1	0.053073	3.4749	3	186	<b>0.017173 *</b>
Residuals	188					

- All variables are statistically significant.

# Linear regression result

Outcome	CD4 <sup>0.33</sup>		CD8 <sup>0.33</sup>		log(RNA)	
	Estimate	Std. Er	Estimate	Std. Er	Estimate	Std. Er
Intercept	<b>6.191 ***</b>	[0.758]	<b>7.718***</b>	[0.639]	<b>8.066 ***</b>	[0.820]
partner without HIV	0.327	[0.821]	<b>1.658*</b>	[0.692]	1.225	[0.929]
Male	-0.784	[0.817]	0.795	[0.689]	<b>3.489***</b>	[0.881]
Interaction	1.185	[0.894]	-0.089	[0.754]	<b>-2.929**</b>	[1.009]
R2	0.063		0.12		0.14	
Observations	271		271		191	



# More problems ...

- Missing values

```
> dim(AIDS)
```

```
[1] 278 6
```

```
> (MI.AIDS<-mi.info(AIDS) )
```

	names	include	order	number.mis	all.mis		type	collinear
1	id	No	NA	0	No	positive-continuous		No
2	sex	Yes	NA	0	No	binary		No
3	type	Yes	NA	0	No	binary		No
4	cd4	Yes	1	<b>3</b>	No	positive-continuous		No
5	cd8	Yes	2	<b>3</b>	No	positive-continuous		No
6	rna	Yes	3	<b>83</b>	No	positive-continuous		No

# Missing data imputation

```
> missings <- which(is.na(AIDS$rna)=="TRUE")
> predictRNA <- predict(lm(rna.tr~cd4.tr+cd8.tr, data=AIDS,
    na.action=na.omit), AIDS[missings, c("cd4.tr", "cd8.tr")])
>
> AIDS$rna.pred <- AIDS$rna.tr
> AIDS$rna.pred[missings] <- predictRNA
>
> summary(m6 <- lm(cbind(cd4.tr, cd8.tr, rna.pred) ~type*sex,
AIDS))
```

# Response rna.pred:

Call: `lm(formula = rna.pred ~ type * sex, data = AIDS)`

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	8.5209	0.6833	12.471	< 2e-16	***
typeDP	1.1009	0.7400	1.488	0.138004	
sexm	2.9329	0.7370	3.979	8.88e-05	***
typeDP:sexm	-2.7149	0.8065	-3.366	0.000873	***

Residual standard error: 2.05 on 271 degrees of freedom

Multiple R-squared: 0.1135, Adjusted R-squared: 0.1037

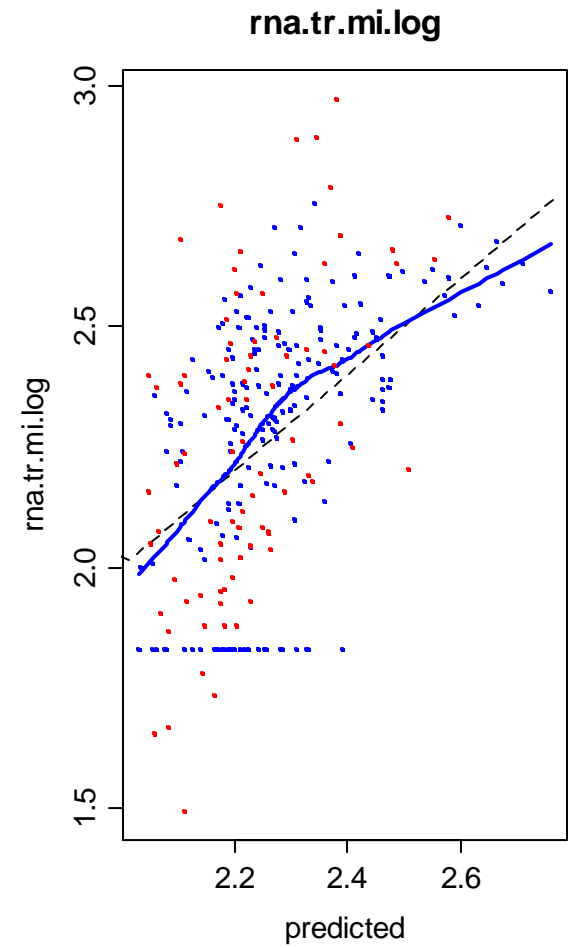
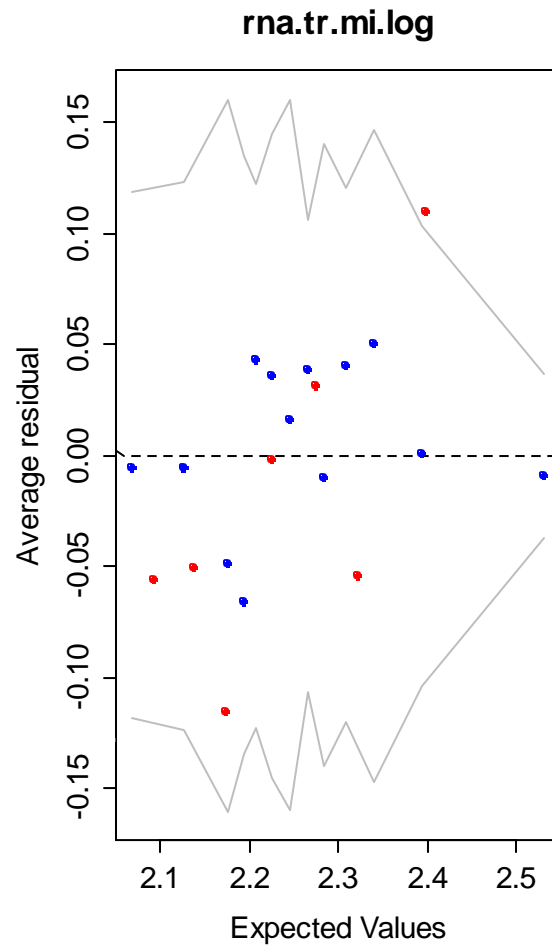
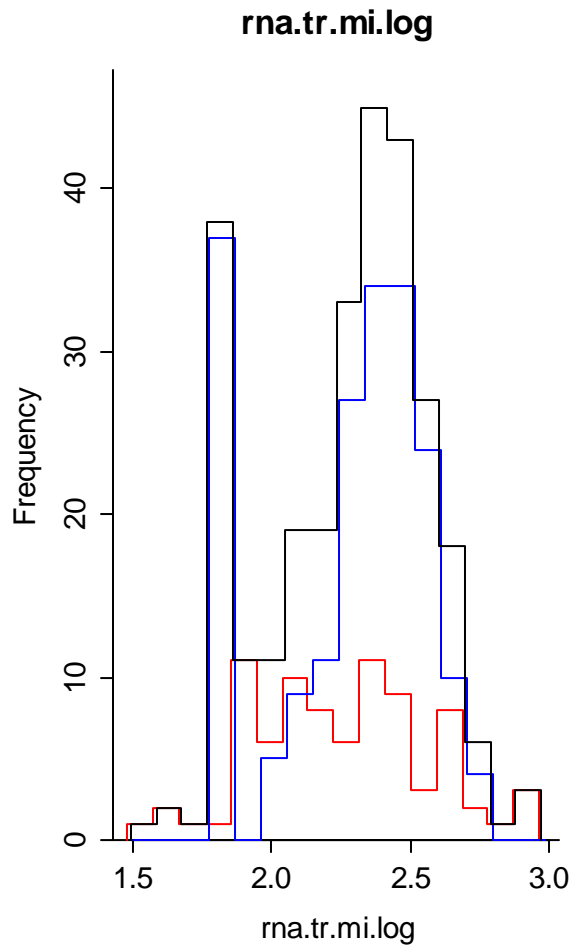
F-statistic: 11.56 on 3 and 271 DF, p-value: 3.718e-07

# “MI” package

```
> AIDS.mi <-AIDS[,c("sex", "type", "cd4.tr", "cd8.tr", "rna.tr")]
> info = mi.info(AIDS.mi)

> AIDS.new = mi.preprocess(AIDS.mi)
> attr(AIDS.new, "mi.info")
> info.upd = update(info, "imp.formula", list("cd4.tr" =
"cd4.tr ~ sex * type + cd8.tr + rna.tr"))
> info.upd = update(info.upd, "imp.formula", list("cd8.tr" =
"cd8.tr ~ sex * type + cd4.tr + rna.tr"))
> info = update(info.upd, "imp.formula", list("rna.tr" =
"rna.tr ~ sex * type + cd4.tr + cd8.tr"))

> AIDS.new = mi.preprocess(AIDS.mi)
> IMP <- mi(AIDS.new, n.iter=50)
```

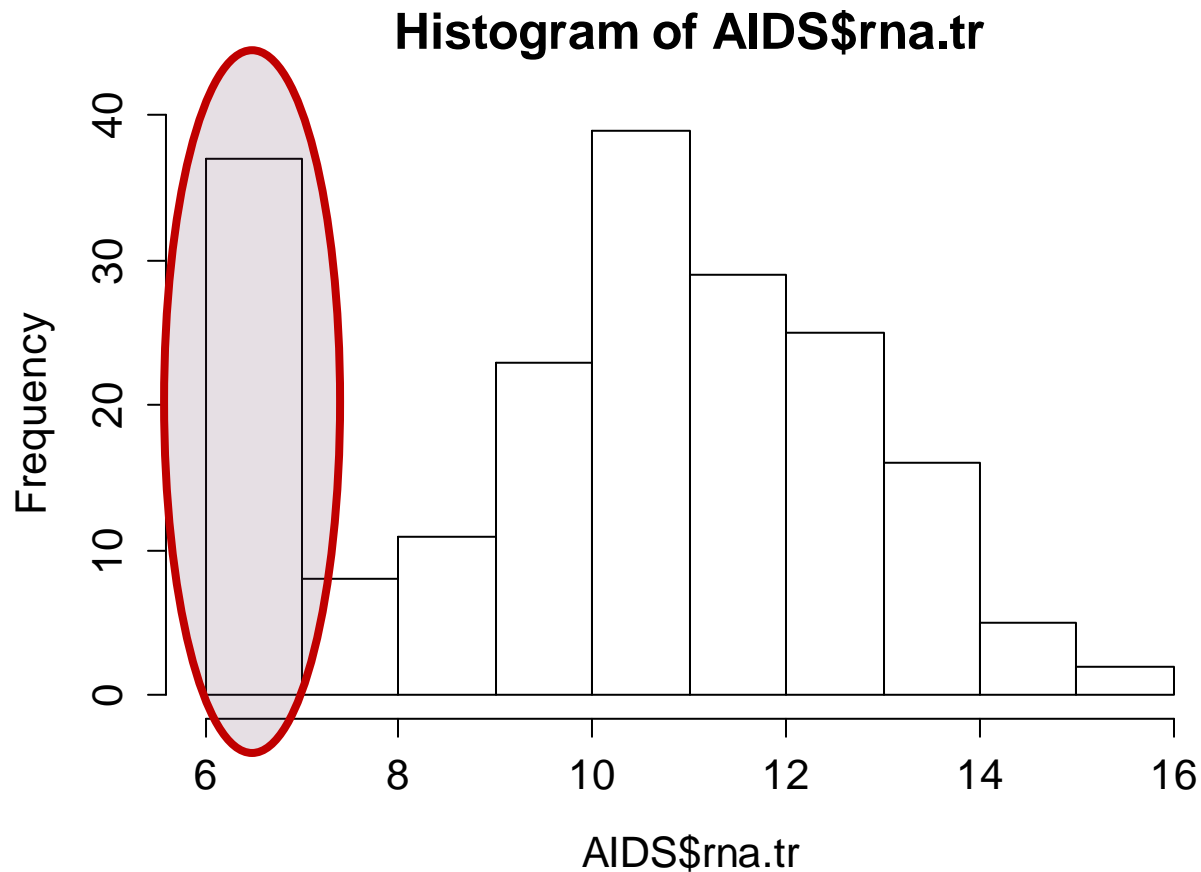


# Linear regression result

Outcome	Log(RNA)		Log(RNA.pred)		log(RNA.mi)	
	Estimate	Std. Er	Estimate	Std. Er	Estimate	Std. Er
Intercept	<b>8.066 ***</b>	[0.820]	<b>8.521***</b>	[0.683]	<b>8.066 ***</b>	[0.824]
partner without HIV	1.225	[0.929]	<b>1.101*</b>	[0.740]	1.225	[0.935]
Male	<b>3.489***</b>	[0.881]	2.933	[0.737]	<b>3.417***</b>	[0.889]
Interaction	<b>-2.929**</b>	[1.009]	-2.7147	[0.807]	<b>-2.856**</b>	[1.018]
R2	0.14		0.10		0.12	
Observations	191		271		274	

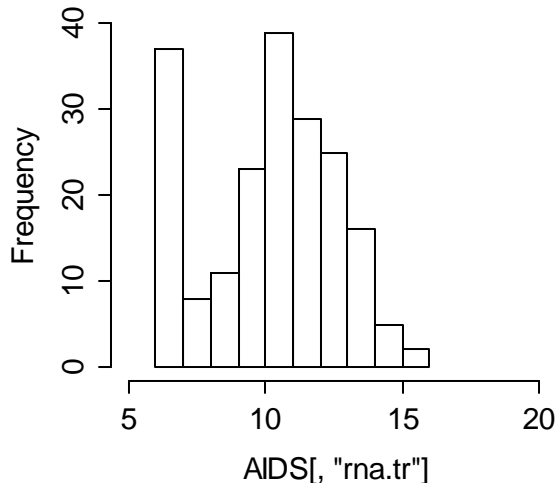
# More problems ...

- Lowest value inflation  
-> against normality assumption

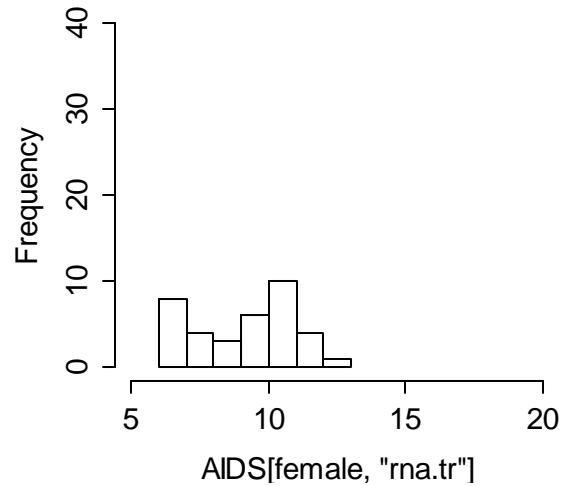


# However,

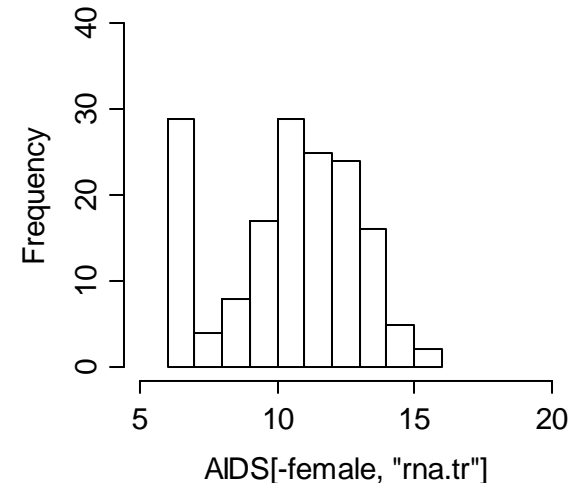
**Histogram of AIDS[, "rna.tr"]**



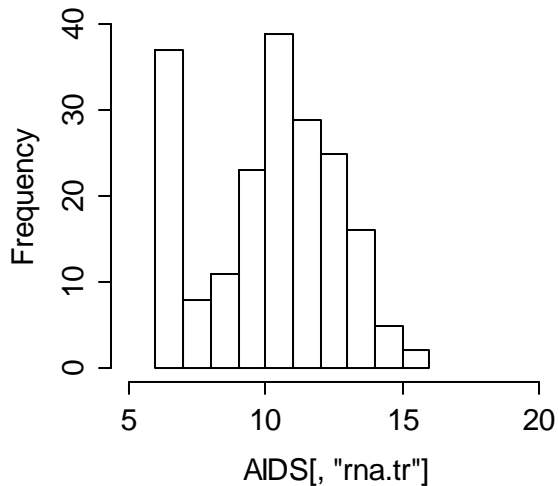
**Histogram of AIDS[female, "rna.tr"]**



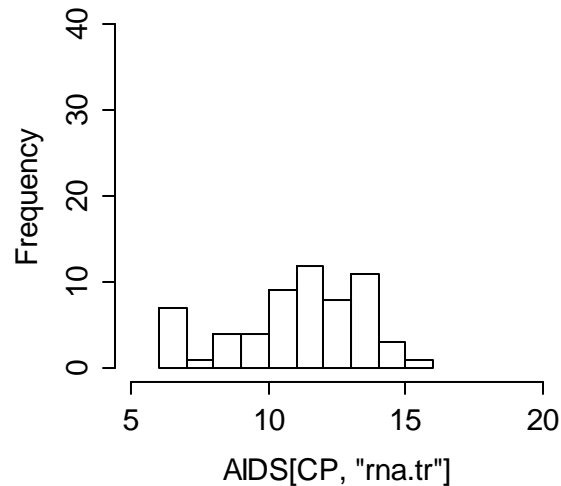
**Histogram of AIDS[-female, "rna.tr"]**



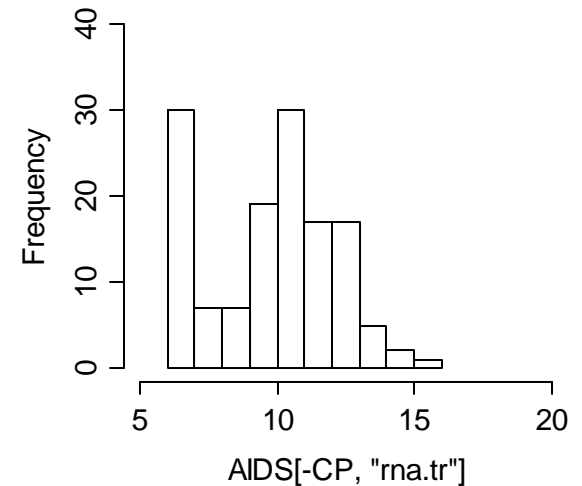
**Histogram of AIDS[, "rna.tr"]**



**Histogram of AIDS[CP, "rna.tr"]**



**Histogram of AIDS[-CP, "rna.tr"]**



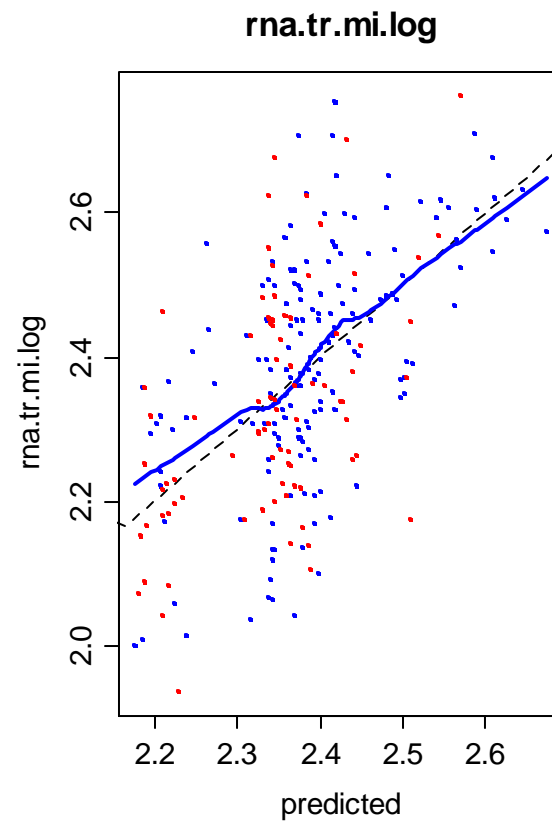
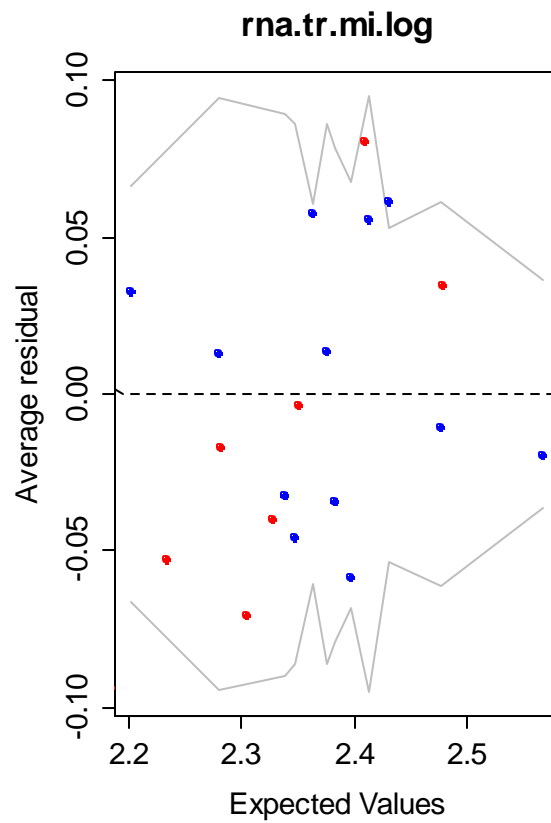
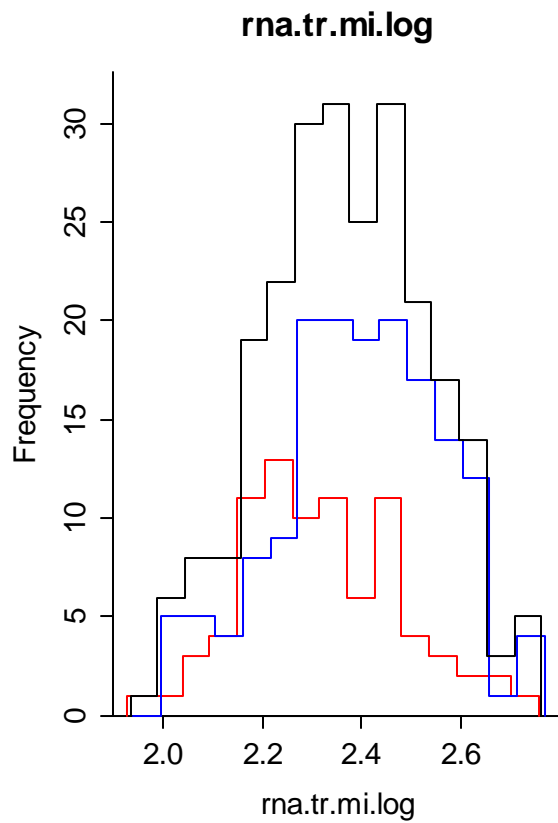


# However,

- The lowest values seem to be distributed randomly over each IV.
- The hypothesis, that their proportions are same, could not be rejected.
- It seems fair that we use data without those lowest points.

# Comparison of results

Outcome	Log(RNA)		Log(RNA.pred)		log(RNA.mi)	
	Estimate	Std. Er	Estimate	Std. Er	Estimate	Std. Er
<b>Intercept</b>	<b>9.177***</b>	[0.718]	<b>9.686***</b>	[0.566]	<b>9.014***</b>	[0.636]
<b>partner without HIV</b>	0.613	[0.799]	0.744	[0.601]	0.540	[0.675]
<b>Male</b>	<b>2.793***</b>	[0.764]	<b>2.225***</b>	[0.599]	<b>3.038***</b>	[0.670]
<b>Interaction</b>	<b>-1.608</b>	[0.862]	<b>-1.783**</b>	[0.643]	<b>-1.697*</b>	[0.721]
<b>R2</b>	0.14		0.10		0.12	
<b>Observations</b>	191		271		274	



# Part IV: Conclusion

# Conclusion

- Goal : to check whether an AIDS carrier partner influences a patient's progression of the HIV virus.
- IVs : partner's gender and status (HIV or not)
- DVs : three different blood measurements (CD4, CD8, and RNA)
- We applied MANOVA, multiple regression and missing data imputation.

- A partner without HIV has a positive effect on CD8.
- The partner status is not statistically significant for CD4 or RNA.
- Males that have a partner without HIV have lower values for RNA (i.e. healthier response) with respect to males that have a partner with HIV.