

# The problems encountered during microarray data analysis

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UP Poznań

Październik 4, 2009

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- One microarray consists of: experimental probe - RNA sample from a patient or a healthy person and control probe - RNA isolated from cell line HL60 (a subtype of AML)

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- Acute myeloid leukemia project
- One microarray consists of: experimental probe - RNA sample from a patient or a healthy person and control probe - RNA isolated from cell line HL60 (a subtype of AML)
- 86 hybridization: 1-2 HL60 versus Control, 3-68 HL60 versus Leukemia, 69-86 HL60 versus Control

# Gpr file from GenePix for AML experiment.

Block	Column	Row	Name	Flags
1	1	1	ERG-Operon	100
1	2	1	ERG-Operon	100
1	3	1	ERG-Operon	100
1	4	1	FLT3-Operon	-50
1	5	1	FLT3-Operon	-50
1	6	1	FLT3-Operon	-50
1	7	1	GAPDHS-Operon	-50
1	8	1	GAPDHS-Operon	-50
1	9	1	GAPDHS-Operon	-50

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- The last column gives us the specified knowledge which weights should be given to spots
- For flags less than the cutoff value we give weights equal 0 and 1 otherwise
- We choose cutoff=-50 to downweight bad or absent spots

# Problem Number One

## Problem

How to calculate the mean intensity for each gene taking into consideration the weight of the spot?

# Problem Number One

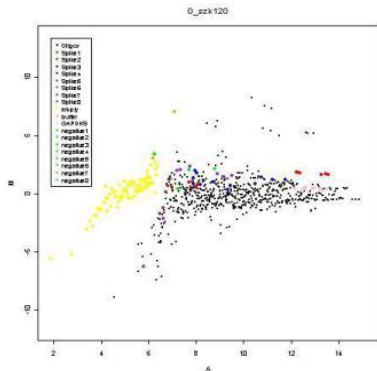
## Problem

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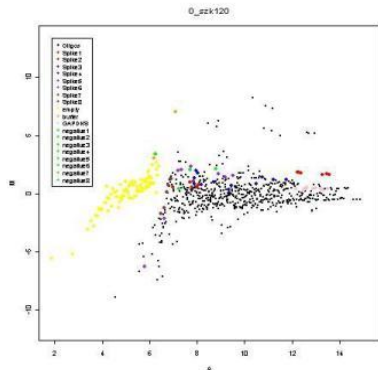
## R code

```
> # MA.A - data after normalization  
> Mean_intensity <- avedups(MA.A, ndups=3,  
weights=MA.A$weights)
```

# MA plots: before and after using avedups function

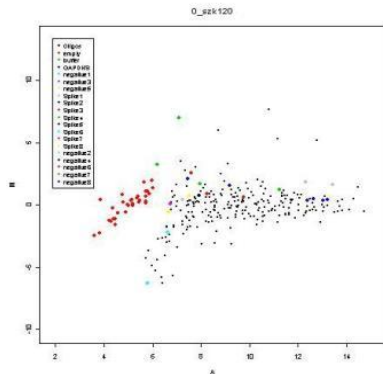
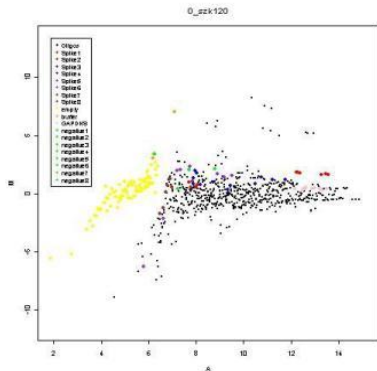


# MA plots: before and after using `avedups` function



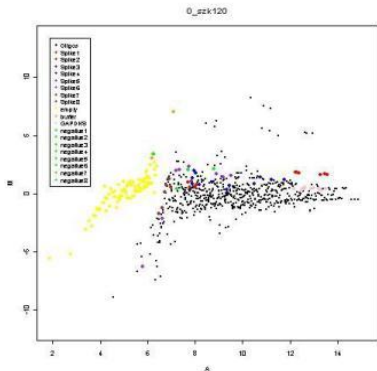
before using `avedups` function

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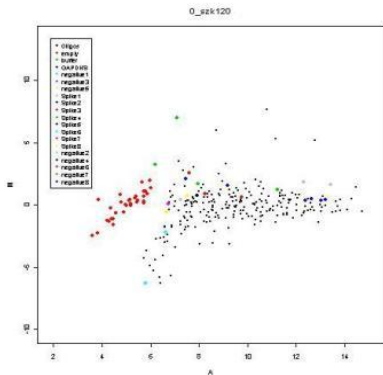


before using `avedups` function

# MA plots: before and after using `avedups` function



before using `avedups` function



after using `avedups` function

# Problem Number Two

## Question

Which genes are over(under)expressed comparing leukemia and control probe?



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- fc-statistics

## DEDS package

Yuanyuan Xiao and Yee Hwa Yang

April 21, 2009

University of California

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```
deds.stat.linkC(X, L, B, tests = c("t", "fc", "sam", "...") )
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- X: A matrix, in the case of gene expression data, rows correspond to N genes and columns to p mRNA samples

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```

- X: A matrix, in the case of gene expression data, rows correspond to N genes and columns to p mRNA samples
- L: A vector of integers corresponding to observation (column) class labels
- B: The number of permutations

## R code

```
> library(DEDS)
> # from targets file 0-control, 1-leukemia
> L<-rep(c(0,1,0),c(2,66,18))
> data<-as.matrix(Mean_intesity)
> d <- deds.stat.linkC(data, L, B=200)
> # for the comparisons between the 3 statistics
> t_genes<-topgenes(d,number=50,Mean_intesity$genes$Name,
+ sort.by="t")
> fc_genes<-topgenes(d,number=50,Mean_intesity$genes$Name,
+ sort.by="fc")
> sam_genes<-topgenes(d,number=50,Mean_intesity$genes$Name,
+ sort.by="sam")
```

# Data from DEDS...

data1	data2
H200011980-NM_006043	H200011164-NM_002317
opHsV0400006953--	H300005238-XM_375664;NM_024762
H300016130-NM_138576	H300008172-NM_006476
opHsV0400005878--	opHsV0400005401--
opHsV0400012693--	opHsV0400006577-NM_181302;NM_144574;
opHsV0400013392--	opHsV0400008010-XM_373962
H300010310--	opHsV0400008839-NM_207355;NM_174981;
opHsV0400005020--	opHsV0400009215-XM_497555
opHsV0400006947--	opHsV0400009537-XM_498325
opHsV0400008803-XM_496095	opHsV0400010719--
opHsV0400011787--	opHsV0400011041--
H300003153-NM_007191	H300003254-NM_014220
opHsV0400000577-NM_153329	H300006844-NM_003295
opHsV0400002475-NM_001010848	H300007739--
opHsV0400005490--	H300018967-NM_006718;NM_002656
opHsV0400007041--	H300022101-NM_022900
opHsV0400007394-XM_292810	opHsV0400010766--
opHsV0400013409--	opHsV0400012663--
H200011667-NM_017907	HumV4con_1-K13-H200012219-NM_000967
H300004333--	H300019333-NM_194463;NM_024539
H300007176--	H300009840-NM_153668
H300007217--	H300019956--
H200013033--	H300020878-NM_005214
H300003287-NM_001001923	H300022082-NM_020357
H300008591-NM_033312	opHsV0400003586--
H300017062-NM_006615	H200000348-NM_000133
H200001066-NM_001006643;NM_001006641;	H200000377-NM_021912;NM_000814
H200008128-NM_014633	H200003989-NM_004236
H200011722-NM_016310	H200011709-NM_003472
H300007333-XM_497715	H200011791-NM_172177;NM_172178;
H300011810-NM_007130	H200016772-NM_002748
H300019192-NM_001297	H200017794-NM_020357
opHsV0400002318-NM_198530;NM_001008529;	H200019486-NM_019063
opHsV0400005702--	H200020562-XM_057296
H300006283-XM_376233	H300002047-NM_015384;NM_133433

# Venn diagram

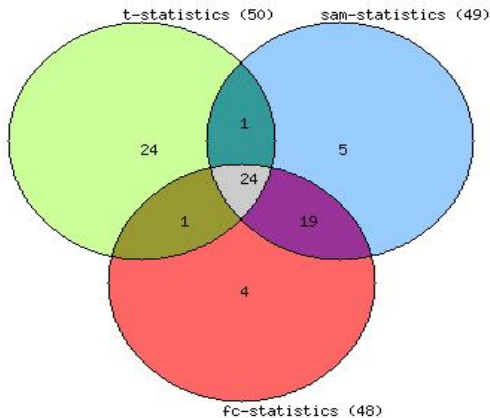
## R code

```
> w<-c(data1,data2)
+ hm<-duplicated(w)
```

# Venn diagram

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```



# Another example

Block	Column	Row	Name	ID
1	1	1	Dye Marker	97: D-01 Dye Marker
1	2	1	H200000001-NM_001885	01-D01-H200000498-ENSG00000109846
1	3	1	Buffer	96: D-01 Buffer
1	4	1	H200000511-NM_030984;NM_001061	01-D13-H200000511-ENSG00000059377
1	5	1	H200000542-NM_005658	01-H01-H200000542-ENSG00000056558
1	6	1	H200000008-NM_005041	01-H13-H200000557-ENSG00000180644
1	7	1	H200000577-NM_000073	01-L01-H200000577-ENSG00000180654
1	8	1	H200000583-NM_003385	01-L13-H200000583-ENSG00000163032
1	9	1	H200000011-NM_006080	01-P01-H200000613-ENSG00000075213

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GPR data

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384_number	384_position	oligo_id	oligo_seq	gene_id	transcript_id	gene_symbol
1	A03	H200000001	TGGGGAGAA	ENSG0000019	ENST0000028	INAT2
1	A05	H200000005	GAAGGCTCT	ENSG0000009	ENST0000020	TGM1
1	A07	H200000006	ATGGGTACA	ENSG0000006	ENST0000038	FECH
1	A09	H200000007	TATGGAGAT	ENSG0000017	ENST0000038	GLDC
1	A11	H200000008	GTCATCTCT	ENSG0000014	ENST0000027	MS4A2
1	A13	H200000010	CATGGAGGA	ENSG0000017	ENST0000038	Q6FG55_HUMAN
1	A15	H200000011	GAACAGGAG	ENSG0000007	ENST0000026	ACAT1
1	A17	H200000014	GTGCTGTGG	ENSG0000018	ENST0000037	PTAFR

GPR data



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GPR data

384_number	384_position	oligo_id	oligo_seq	gene_id	transcript_id	gene_symbol
1	A03	H200000001	TGGGGAGAA	ENSG0000018	ENST0000028	INAT2
1	A05	H200000005	GAAGGCTCT	ENSG0000009	ENST0000020	TGM1
1	A07	H200000006	ATGGGTACA	ENSG0000008	ENST0000038	FECH
1	A09	H200000007	TATGGAGAT	ENSG0000017	ENST0000038	GLDC
1	A11	H200000008	GTCATCTCT	ENSG0000014	ENST0000027	MS4A2
1	A13	H200000010	CATGGAGGA	ENSG0000017	ENST0000038	Q6FG55_HUMAN
1	A15	H200000011	GAACAGGAG	ENSG0000007	ENST0000026	ACAT1
1	A17	H200000014	GTGCTGTGG	ENSG0000018	ENST0000037	PTAFR

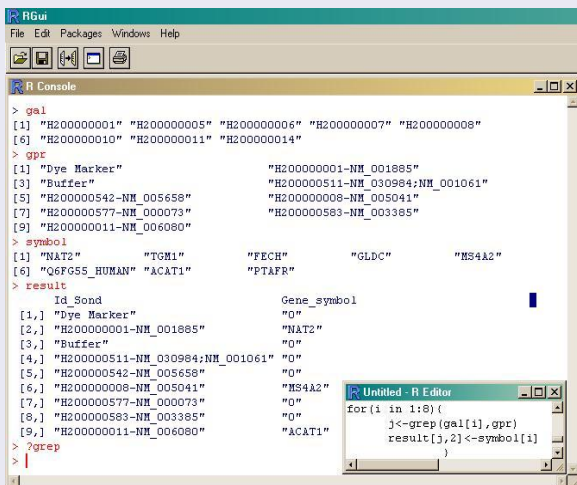
GAL data

## R code

```
> gal<-read.table("gal.csv",dec="," , sep=";")
> gpr<-read.table("gpr.csv",dec="," , sep=";")
> gal<-gal[,3]
> gal<-as.character(gal)
> gpr<-gpr[,4]
> gpr<-as.character(gpr)
> symbol<-gal[,9]
> symbol<-as.character(symbol)
> result<-matrix(0,length(gpr),2)
> result[,1]<-gpr
> colnames(result)<-c("Sonda", "Gen_symbol")
```

# Data after grep function

Finally we obtain id sond in the first column and the gene symbol in the second



```
RGui
File Edit Packages Windows Help
[Icons]

R Console
> gal
[1] "H200000001" "H200000005" "H200000006" "H200000007" "H200000008"
[6] "H200000010" "H200000011" "H200000014"
> gpr
[1] "Dye Marker" "H200000001-NM_001885"
[3] "Buffer" "H200000511-NM_030984;NM_001061"
[5] "H200000542-NM_005658" "H200000008-NM_005041"
[7] "H200000577-NM_000073" "H200000583-NM_003385"
[9] "H200000011-NM_006080"
> symbol
[1] "NAT2" "TGM1" "FECH" "GLDC" "MS4A2"
[6] "Q6FG55_HUMAN" "ACAT1" "PTAFR"
> result
  Id_Sond Gene_symbol
[1,] "Dye Marker" "0"
[2,] "H200000001-NM_001885" "NAT2"
[3,] "Buffer" "0"
[4,] "H200000511-NM_030984;NM_001061" "0"
[5,] "H200000542-NM_005658" "0"
[6,] "H200000008-NM_005041" "MS4A2"
[7,] "H200000577-NM_000073" "0"
[8,] "H200000583-NM_003385" "0"
[9,] "H200000011-NM_006080" "ACAT1"
> ?grep
> |

R Untitled - R Editor
for(i in 1:8){
  j<-grep(gal[i],gpr)
  result[j,2]<-symbol[i]
}
```

:-)