The CGDS-R library

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1 Introduction

This package provides a basic set of R functions for querying the Cancer Genomic Data Server (CGDS) hosted by the Computational Biology Center (cBio) at the Memorial Sloan-Kettering Cancer Center (MSKCC). This service is a part of the cBio Cancer Genomics Portal, http://www.cbioportal.org/.

In summary, the library can issue the following types of queries:

- getCancerStudies(): What cancer studies are hosted on the server? For example, TCGA glioblastoma or TCGA ovarian cancer.
- getGeneticProfiles(): What genetic profile types are available for cancer study X? For example, mRNA expression or copy number alterations.
- getCaseLists(): what case sets are available for cancer study X? For example, all samples or only samples corresponding to a given cancer subtype.

- getProfileData(): Retrieve slices of genomic data. For example, a client can retrieve all mutation data for PTEN and EGFR in TCGA glioblastoma.
- getClinicalData(): Retrieve clinical data (e.g. patient survival time and age) for a given cancer study and list of cases.

Each of these functions will be briefly described in the following sections. The last part of this document includes some concrete examples of how to access and plot the data.

The purpose of this document is to give the reader a quick overview of the cgdsr package. Please refer to the corresponding R manual pages for a more detailed explanation of arguments and output for each function.

2 The CGDS R interface

2.1 CGDS(): Create a CGDS connection object

Initially, we will establish a connection to the public CGDS server hosted by Memorial Sloan-Kettering Cancer Center. The function for creating a CGDS connection object requires the URL of the CGDS server service, in this case http://www.cbioportal.org/, as an argument.

```
> library(cgdsr)
> # Create CGDS object
> mycgds = CGDS("http://www.cbioportal.org/")
```

The variable mycgds is now a CGDS connection object pointing at the URL for the public CGDS server. This connection object must be included as an argument to all subsequent interface calls. Optionally, we can now perform a set of simple tests of the data returned from the CGDS connection object using the test function:

```
> # Test the CGDS endpoint URL using a few simple API tests
> test(mycgds)

getCancerStudies... OK
getCaseLists (1/2) ... OK
getCaseLists (2/2) ... OK
getGeneticProfiles (1/2) ... OK
```

getClinicalData (1/1) ... OK
getProfileData (1/6) ... OK
getProfileData (2/6) ... OK
getProfileData (3/6) ... OK
getProfileData (4/6) ... OK
getProfileData (5/6) ... OK

getProfileData (6/6) ...

getGeneticProfiles (2/2) ...

2.2 getCancerStudies(): Retrieve a set of available cancer studies

Having created a CGDS connection object, we can now retrieve a data frame with available cancer studies using the getCancerStudies function:

> # Get list of cancer studies at server
> getCancerStudies(mycgds)[,c(1,2)]

```
cancer_study_id
1
                        paac_jhu_2014
2
                        laml_tcga_pub
3
                            laml_tcga
4
                        acyc_fmi_2014
5
                        acyc_mda_2015
6
                      acyc_mskcc_2013
7
                     acyc_sanger_2013
8
                      acbc_mskcc_2015
9
                             acc_tcga
                       ampca_bcm_2016
10
               blca_mskcc_solit_2014
11
               blca_mskcc_solit_2012
12
        blca_plasmacytoid_mskcc_2016
13
                             blca_bgi
14
15
             blca_dfarber_mskcc_2014
16
                        blca_tcga_pub
17
                            blca_tcga
18
                             lgg_tcga
19
                        brca_metabric
20
                           brca_bccrc
21
                           brca_broad
22
                          brca_sanger
23
                    brca_tcga_pub2015
24
                        brca_tcga_pub
25
                            brca_tcga
26
           brca_bccrc_xenograft_2014
27
                 cellline_ccle_broad
28
                            cesc_tcga
29
                       chol_nccs_2013
                        chol_nus_2012
30
31
                            chol_tcga
32
                      lcll_broad_2013
33
                       cll_iuopa_2015
34
                    ccrcc_utokyo_2013
35
                   coadread_dfci_2016
36
                   coadread_genentech
37
                    coadread_tcga_pub
38
                        coadread_tcga
39
                       coadread_mskcc
                   ctcl_columbia_2015
40
41
                    cscc_dfarber_2015
```

42	pact_jhu_2011
43	desm_broad_2015
44	dlbc_broad_2012
45	esca_broad
46	esca_tcga
47	escc_icgc
48	escc_ucla_2014
49	es_iocurie_2014
50	gbc_shanghai_2014
51	egc_tmucih_2015
52	nsclc_unito_2016
53	prad_cpcg_2017
54	gct_msk_2016
55	gbm_tcga_pub2013
56	gbm_tcga_pub
57	gbm_tcga
58	hnsc_broad
59	hnsc_jhu
60	hnsc_tcga_pub
61	hnsc_tcga
62	liad_inserm_fr_2014
63	hcc_inserm_fr_2015
64	all_stjude_2013
65	all_stjude_2015
66	panet_shanghai_2013
67	chol_jhu_2013
68	kich_tcga_pub
69	kich_tcga
70	kirc_bgi
71	kirc_tcga_pub
72	kirc_tcga
73	kirp_tcga
74	lihc_amc_prv
75	lihc_riken
76	lihc_tcga
77	lgg_ucsf_2014
78	luad_broad
79	luad_mskcc_2015
80	luad_tcga_pub
81	luad_tcga
82	luad_tsp
83	lusc_tcga_pub
84	lusc_tcga
85	dlbc_tcga
86	msk_impact_2017
87	lung_msk_2017
88	prad_mskcc_2017
89	mpnst_mskcc
90	plmeso_nyu_2015
91	mcl_idibips_2013
01	mcr_rarbips_zoro

```
92
                       mbl_broad_2012
93
                             mbl_icgc
94
                             mbl_pcgp
95
                   mbl_sickkids_2016
96
                  skcm_broad_dfarber
97
                      lgggbm_tcga_pub
98
                            meso_tcga
99
                       prad_su2c_2015
100
                             mm_broad
101
                       ccrcc_irc_2014
                       brca_igr_2015
102
103
                       mds_tokyo_2011
104
                       cellline_nci60
105
                         odg_msk_2017
106
                      npc_nusingapore
107
                         nbl_amc_2012
108
                   nbl_ucologne_2015
109
                        nepc_wcm_2016
110
          skcm_vanderbilt_mskcc_2015
111
                hnsc_mdanderson_2013
112
                          ov_tcga_pub
113
                              ov_tcga
114
                 mel_tsam_liang_2017
115
               nsclc_tcga_broad_2016
116
                            paad_icgc
117
                   paad_qcmg_uq_2016
118
                            paad_tcga
119
                       paad_utsw_2015
120
                       panet_jhu_2011
121
                        thca_tcga_pub
122
               es_dfarber_broad_2014
123
                            pcpg_tcga
124
                  thyroid_mskcc_2016
125
                      pcnsl_mayo_2015
126
                     prad_broad_2013
127
                           prad_broad
128
                           prad_fhcrc
                           prad_mskcc
129
130
                        prad_tcga_pub
131
                            prad_tcga
132
                      prad_mskcc_2014
133 prad_mskcc_cheny1_organoids_2014
134
                            prad_mich
135
                       hnc_mskcc_2016
136
               nccrcc_genentech_2014
137
                         rms_nih_2014
138
                           sarc_mskcc
139
                            sarc_tcga
140
                           skcm_broad
141
                            skcm_tcga
```

142	skcm_yale	
143	scco_mskcc	
144	sclc_clcgp	
145	sclc_jhu	
146	sclc_ucologne_2015	
147	stad_pfizer_uhongkong	
148	stad_tcga_pub	
149	stad_tcga	
150	stad_utokyo	
151	stad_uhongkong	
152	stes_tcga_pub	
153	egc_msk_2017	
154	urcc_mskcc_2016	
155	crc_msk_2018	
156	utuc_mskcc_2013	
157	tgct_tcga	
158	brca_mbcproject_wagle_2017	
159	tet_nci_2014	
160	thym_tcga	
161	thca_tcga	
162	ucs_jhu_2014	
163	ucs_tcga	
164	ucec_tcga_pub	
165	ucec_tcga	
166	uvm_tcga	
167	panet_arcnet_2017	
168	skcm_ucla_2016	
169	past_dkfz_heidelberg_2013	
1		Acinar Cell Carcinoma of the Pancreas (Joh
2		Acute Myeloid :
3		Acute Myeloid Le
4		Adenoid Cystic Carcinoma (F.
5		Adenoid Cystic Carcinoma (1
6		Adenoid Cystic Carcino
7		Adenoid Cystic Carcin
8		Adenoid Cystic Carcinoma of the Brea
9		Adrenocortical Car
10		Ampullary Carcinoma (Baylor College of Me
11		Bladder Can
12		Bladde
13		Bladder Cancer, Plasmacytoid Varia
14		Bladder Urothelial Carci:
15		Bladder Urothelial Carcinoma (Dana Farber & 1
16		Bladder Urothelial Car
17		Bladder Urothelial Car
18		Brain Lower Grade
19		Breast Cancer (METABRIC, Natu
20		Breast Invasive Carcinoma (Brit
21		Breast Invasive Carc

Chronic Lymphocytic I	33
Clear Cell Renal Cell Carcino	34
Colorectal Adenocarcing	35
Colorectal Adenocarcin	36
Colorectal Adenoc	37
Colorectal Adenoc	38
Colorectal Adenocarcinoma Triple	39
Cutaneous T Cell Lymphoma	40
Cutaneous squamous cell carcinoma	41
Cystic Tumor of the Pancre	42
Desmoplastic Melanoma (Broa	43
Diffuse Large B-Cell	44
Esophageal Adenocarci	45
Esophageal C	46
Esophageal Squamous Cell C	47
Esophageal Squamous Cell Card	48
Ewing Sarcoma (Institu	49
Gallbladder Carcinom	50
Gastric Adenoc	51
Genetic Characterization of NSCLC young adult patients (University	52
Genomic Hallmarks of Prostate Adenocarci	53
Genomic Profile of Patients with Advanced Germ	54
Gl	55
Glic	56
Glioblastoma Mu	57
Head and Neck Squamous Cell Car	58
Head and Neck Squamous Cell Carcinoma (59
Head and Neck Squamous Cell C	60
Head and Neck Squamous Cell C	61
Hepatocellular Adenom	62
Hepatocellular Carcino	63
Hypodiploid Acute Lymphoid Leukem	64
Infant MLL-Rearranged Acute Lymphoblastic Leukem	65
Insulinoma	66
Intrahepatic Cholangiocarcinoma (Johns Hopkins	67
Kidney Chromoph	68
Kidney Chr	69
Kidney Renal Clear Cell Car	70
Kidney Renal Clear Cell C	71
v	

Breast Invasive Carci

Breast Invasive C

Cholangiocar

Breast Invasive Car

Breast Invasive Car

Chronic Lymphocytic L

Breast cancer patient xenografts (Brit

Cholangiocarcinoma (National Cancer Centre of

Cholangiocarcinoma (National University of

Cervical Squamous Cell Carcinoma and Endocervical Adenocar

Cancer Cell Line Encyclopedia (No

22

23

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27

28 29

30

31

Liver Hepatocellular Carcii
Liver Hepatocellular Ca
Low-Grade G
Lung Adeno
Lung
Lung Adenoca
Lung Adenoca
Lung Adeno
Lung Squamous Cell Ca
Lung Squamous Cell Ca
Lymphoid Neoplasm Diffuse Large B-cell N
MSK-IMPACT Clinical Sequencing (
MSK-IMPACT Clinical Sequencing Cohort for Non-Small Cell Cancer
MSK-IMPACT Clinical Sequencing Cohort in Prostate Cancer (MSK,
Malignant Peripheral Nerve Sheath To
Malignant Pleural Mesothe
Mantle Cell Ly
Medullob
Medullo
Medullo
Medulloblas
Melanoma (Bro
Merged Cohort of Lo
Meso
Metastatic Prostate Cancer, SU2C/PCF Dream Team
Multiple Myelor
Multiregion Sequencing of Clear Cell Renal Cell Card
2 Mutational profiles of metastatic
3 Myelody:
NCI-60 Cell L
NGS in Anaplastic Oligodendroglioma and Anaplastic Oligoastrocytomas tu
Nasopharyngeal Carcinoma
7 Neuroblastoma
Neuroblas ⁻
9 Neuroendocrine Prostate Cancer
O Next generation sequencing (NGS) of pre-treatment metastatic melanoma samples (MSK,
Oral Squamous Cell Carcinoma (MD A
2 Ovarian Serous Cystadenoc
Ovarian Serous Cystadenoc
Paired-exome sequencing of acral mela
5 Pan-Lung C
6 Pancreatic Adenoc
7 Pancreatic Adenoc
Pancreatic Adenocation
9 Pancreatic Ca
Pancreatic Neuroendocrine Tumors (Johns Hopkin
Papillary Thyroid
- aprilary injusta

Kidney Renal Clear Cell Car

Liver Hepatocellular Carcino

Kidney Renal Papillary Cell Car Liver Hepatocellular Carcin

72

73

74 75

Pediatric Ewing Sarcon
3 Pheochromocytoma and Parag
4 Poorly-Differentiated and Anaplastic Thyro
5 Primary Central Nervous System Lymphoma (Mayo C
6 Prostate Adenocarcinon
7 Prostate Adenocarcinoma (Br
8 Prostate Adenocarcinoma (Fred F
9 Prostate Adenocarcing
O Prostate Ader
1 Prostate Adenoc
2 Prostate Adenocarcinoma
3 Prostate Adenocarcinoma
4 Prostate Adenocarcinoma, Metast
5 Recurrent and Metastatic Head & Neck Car
6 Renal Non-Clear Cell Carcinoma
7 Rhabdomyosarco
8 Sarcoma (
9
O Skin Cutaneous
1 Skin Cutaneous
Skin Cutaneous Mel
3 Small Cell Carcinoma of the C
4 Small Cell Lung Ca
5 Small Cell Lung Cancer (Jo
6 Small Cell Lung Car
7 Stomach Adenocarcinoma (Pfi
8 Stomach Adenoc
9 Stomach Adenoc
O Stomach Adenocarcing
1 Stomach Adenocar
2 TCGA data for Esophagus-Stomach
3 Targeted Sequencing of 341 samples from metastatic esophagogastric cancer patients
4 Targeted gene sequencing in 62 high-grade primary Unclassified Renal Cell
5 Targeted sequencing of 1134 samples from metastatic colorectal cancer sam
6 Targeted sequencing of upper tract urothelial cancer
7 Testicular Germ Cel
8 The Metastatic Breast Cancer Project
9 Thymic Epithelial
0
1 Thyroid (
2 Uterine Carcinosarcoma (Johns Hopkins
3 Uterine Carcin
4 Uterine Corpus Endometrial (
5 Uterine Corpus Endometrial C
6 Uveal
7 Whole-Genome Sequencing of Pancreatic Neuroend
Whole-evens sequences (MES) of pretreatment melan

Whole-exome sequences (WES) of pretreatment melanom

Whole-genome sequencing of pilocytic astrocytoma

Here we are only showing the first two columns, the cancer study ID and

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short name, of the result data frame. There is also a third column, a longer description of the cancer study. The cancer study ID must be used in subsequent interface calls to retrieve case lists and genetic data profiles (see below).

2.3 getGeneticProfiles(): Retrieve genetic data profiles for a specific cancer study

This function queries the CGDS API and returns the available genetic profiles, e.g. mutation or copy number profiles, stored about a specific cancer study. Below we list the current genetic profiles for the TCGA glioblastoma cancer study:

> getGeneticProfiles(mycgds, 'gbm_tcga')[,c(1:2)]

```
genetic_profile_id
                              gbm_tcga_rppa
2
                      gbm_tcga_rppa_Zscores
3
                            gbm_tcga_gistic
4
                         gbm_tcga_mrna_U133
5
                gbm_tcga_mrna_U133_Zscores
6
                              gbm_tcga_mrna
7
              gbm_tcga_mrna_median_Zscores
8
                  gbm_tcga_rna_seq_v2_mrna
   gbm_tcga_rna_seq_v2_mrna_median_Zscores
9
10
                        gbm_tcga_linear_CNA
                 gbm_tcga_methylation_hm27
11
12
                gbm_tcga_methylation_hm450
13
                         gbm_tcga_mutations
                               genetic_profile_name
                          Protein expression (RPPA)
1
2
                Protein expression Z-scores (RPPA)
3
      Putative copy-number alterations from GISTIC
4
            mRNA expression (U133 microarray only)
5
   mRNA Expression z-Scores (U133 microarray only)
6
                      mRNA expression (microarray)
7
             mRNA Expression z-Scores (microarray)
8
                 mRNA expression (RNA Seq V2 RSEM)
9
        mRNA Expression z-Scores (RNA Seq V2 RSEM)
10
                Relative linear copy-number values
11
                                 Methylation (HM27)
12
                                Methylation (HM450)
                                          Mutations
13
```

Here we are only listing the first two columns, genetic profile ID and short name, of the resulting data frame. Please refer to the R manual pages for a more extended specification of the arguments and output.

2.4 getCaseLists(): Retrieve case lists for a specific cancer study

This function queries the CGDS API and returns available case lists for a specific cancer study. For example, within a particular study, only some cases may have sequence data, and another subset of cases may have been sequenced and treated with a specific therapeutic protocol. Multiple case lists may be associated with each cancer study, and this method enables you to retrieve meta-data regarding all of these case lists. Below we list the current case lists for the TCGA glioblastoma cancer study:

> getCaseLists(mycgds, 'gbm_tcga')[,c(1:2)]

```
case_list_id
       gbm_tcga_3way_complete
1
2
                 gbm_tcga_all
3
           gbm_tcga_sequenced
4
                 gbm_tcga_cna
    gbm_tcga_methylation_hm27
5
6
   gbm_tcga_methylation_hm450
7
                gbm_tcga_mrna
8
     gbm_tcga_rna_seq_v2_mrna
9
           gbm_tcga_mrna_U133
10
                gbm_tcga_rppa
11
              gbm_tcga_cnaseq
                                         case_list_name
                                    All Complete Tumors
1
2
                                              All Tumors
3
                                       Sequenced Tumors
4
                            Tumor Samples with CNA data
5
            Tumor Samples with methylation data (HM27)
6
           Tumor Samples with methylation data (HM450)
7
     Tumor Samples with mRNA data (Agilent microarray)
8
             Tumor Samples with mRNA data (RNA Seq V2)
9
   Tumor Samples with mRNA data (U133 microarray only)
                           Tumor Samples with RPPA data
10
            Tumor Samples with sequencing and CNA data
11
```

Here we are only listing the first two columns, case list ID and short name, of the resulting data frame. Please refer to the R manual pages for a more extended specification of the arguments and output.

2.5 getProfileData(): Retrieve genomic profile data for genes and genetic profiles

The function queries the CGDS API and returns data based on gene(s), genetic profile(s), and a case list. The function only allows specifying a list of genes and a single genetic profile, or oppositely a single gene and a list of genetic profiles. Importantly, the format of the output data frame depends on if a single or a list of genes was specified in the arguments. Below we are retrieving mRNA

expression and copy number alteration genetic profiles for the NF1 gene in all samples of the TCGA glioblastoma cancer study:

> getProfileData(mycgds, "NF1", c("gbm_tcga_gistic","gbm_tcga_mrna"), "gbm_tcga_all")[c(1:

	<pre>gbm_tcga_gistic</pre>	gbm_tcga_mrna
TCGA.02.0001.01	-1	NaN
TCGA.02.0003.01	0	NaN
TCGA.02.0006.01	0	NaN
TCGA.02.0007.01	0	NaN
TCGA.02.0009.01	0	NaN

We are here only showing the first five rows of the data frame. In the next example, we are retrieving mRNA expression data for the MDM2 and MDM4 genes:

> getProfileData(mycgds, c("MDM2", "MDM4"), "gbm_tcga_mrna", "gbm_tcga_all")[c(1:5),]

```
MDM2 MDM4
TCGA.02.0001.01 NaN NaN
TCGA.02.0003.01 NaN NaN
TCGA.02.0006.01 NaN NaN
TCGA.02.0007.01 NaN NaN
TCGA.02.0009.01 NaN NaN
```

We are again only showing the first five rows of the data frame.

2.6 getClinicalData(): Retrieve clinical data for a list of cases

The function queries the CGDS API and returns available clinical data (e.g. patient survival time and age) for a given case list. Results are returned in a data frame with a row for each case and a column for each clinical attribute. The available clinical attributes are:

- overall_survival_months: Overall survival, in months.
- overall_survival_status: Overall survival status, usually indicated as "LIVING" or "DECEASED".
- disease_free_survival_months: Disease free survival, in months.
- disease_free_survival_status: Disease free survival status, usually indicated as "DiseaseFree" or "Recurred/Progressed".
- age_at_diagnosis: Age at diagnosis.

Below we retrieve clinical data for the TCGA ovarian cancer dataset (only first five cases/rows are shown):

```
> getClinicalData(mycgds, "ova_all")[c(1:5),]
```

data frame with 0 columns and 5 rows

3 Examples

3.1 Example 1: Association of NF1 copy number alteration and mRNA expression in glioblastoma

As a simple example, we will generate a plot of the association between copy number alteration (CNA) status and mRNA expression change for the NF1 tumor suprpressor gene in glioblastoma. This plot is very similar to Figure 2b in the TCGA research network paper on glioblastoma (McLendon et al. 2008). The mRNA expression of NF1 has been median adjusted on the gene level (by globally subtracting the median expression level of NF1 across all samples).

> df = getProfileData(mycgds, "NF1", c("gbm_tcga_gistic","gbm_tcga_mrna"), "gbm_tcga_all")
> head(df)

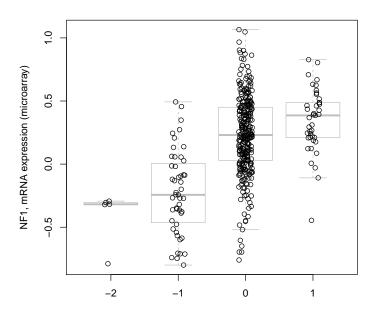
	gbm_tcga_gistic	gbm_tcga_mrna
TCGA.02.0001.01	-1	NaN
TCGA.02.0003.01	0	NaN
TCGA.02.0006.01	0	NaN
TCGA.02.0007.01	0	NaN
TCGA.02.0009.01	0	NaN
TCGA.02.0010.01	0	NaN

> boxplot(df[,2] ~ df[,1], main="NF1 : CNA status vs mRNA expression", xlab="CNA status",
> stripchart(df[,2] ~ df[,1], vertical=T, add=T, method="jitter",pch=1,col='red')

NF1: CNA status vs mRNA expression

Alternatively, the generic cgdsr plot() function can be used to generate a similar plot:

> plot(mycgds, "gbm_tcga", "NF1", c("gbm_tcga_gistic", "gbm_tcga_mrna"), "gbm_tcga_all", sk
[1] TRUE



NF1, Putative copy-number alterations from GISTIC

3.2 Example 2: MDM2 and MDM4 mRNA expression levels in glioblastoma

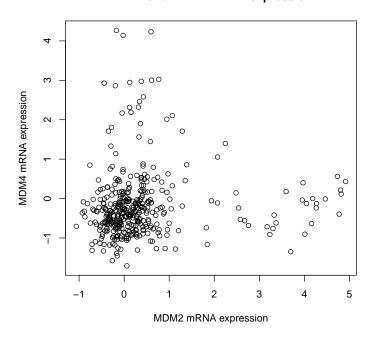
In this example, we evaluate the relationship of MDM2 and MDM4 expression levels in glioblastoma. mRNA expression levels of MDM2 and MDM4 have been median adjusted on the gene level (by globally subtracting the median expression level of the individual gene across all samples).

```
> df = getProfileData(mycgds, c("MDM2","MDM4"), "gbm_tcga_mrna", "gbm_tcga_all")
> head(df)
```

```
MDM2 MDM4
TCGA.02.0001.01
                   {\tt NaN}
                         NaN
TCGA.02.0003.01
TCGA.02.0006.01
                   NaN
                         NaN
TCGA.02.0007.01
                   NaN
                         NaN
TCGA.02.0009.01
                   NaN
                         NaN
TCGA.02.0010.01
                   \mathtt{NaN}
```

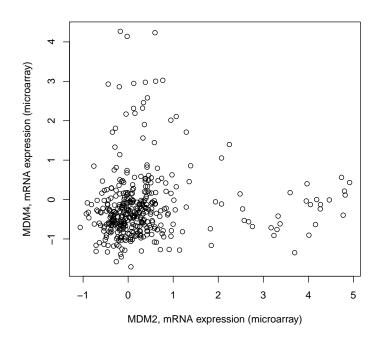
> plot(df, main="MDM2 and MDM4 mRNA expression", xlab="MDM2 mRNA expression", ylab="MDM4 m

MDM2 and MDM4 mRNA expression



Alternatively, the generic $\operatorname{\mathsf{cgdsr}}$ $\operatorname{\mathsf{plot}}()$ function can be used to generate a similar plot:

> plot(mycgds, "gbm_tcga", c("MDM2", "MDM4"), "gbm_tcga_mrna", "gbm_tcga_all")
[1] TRUE



3.3 Example 3: Comparing expression of PTEN in primary and metastatic prostate cancer tumors

In this example we plot the mRNA expression levels of PTEN in primary and metastatic prostate cancer tumors.

```
> df.pri = getProfileData(mycgds, "PTEN", "prad_mskcc_mrna_median_Zscores", "prad_mskcc_pr
> head(df.pri)

PTEN
```

PCA0001 9.467183
PCA0002 9.041528
PCA0003 8.511305
PCA0004 NaN
PCA0005 9.413217
PCA0006 NaN

> df.met = getProfileData(mycgds, "PTEN", "prad_mskcc_mrna_median_Zscores", "prad_mskcc_me
> head(df.met)

PTEN
PCA0182 7.486938
PCA0183 NaN
PCA0184 7.578755
PCA0185 NaN
PCA0186 NaN
PCA0187 8.756132

> boxplot(list(t(df.pri),t(df.met)), main="PTEN expression in primary and metastatic tumor > stripchart(list(t(df.pri),t(df.met)), vertical=T, add=T, method="jitter",pch=1,col='red'

PTEN expression in primary and metastatic tumors

