

```
import numpy as np
import pandas as pd
```

```
train = pd.read_csv('hotel_bookings.csv', nrows=10000)
```

```
train.head()
```

	hotel	is_canceled	lead_time	arrival_date_year
arrival_date_month \				
0	Resort Hotel	0	342	2015
July				
1	Resort Hotel	0	737	2015
July				
2	Resort Hotel	0	7	2015
July				
3	Resort Hotel	0	13	2015
July				
4	Resort Hotel	0	14	2015
July				

	arrival_date_week_number	arrival_date_day_of_month \
0	27	1
1	27	1
2	27	1
3	27	1
4	27	1

	stays_in_weekend_nights	stays_in_week_nights	adults	...
deposit_type \				
0	0	0	2	... No
Deposit				
1	0	0	2	... No
Deposit				
2	0	1	1	... No
Deposit				
3	0	1	1	... No
Deposit				
4	0	2	2	... No
Deposit				

	agent	company	days_in_waiting_list	customer_type	adr \
0	NaN	NaN	0	Transient	0.0
1	NaN	NaN	0	Transient	0.0
2	NaN	NaN	0	Transient	75.0
3	304.0	NaN	0	Transient	75.0
4	240.0	NaN	0	Transient	98.0

	required_car_parking_spaces	total_of_special_requests
reservation_status \		
0	0	0

Check-Out		
1	0	0
Check-Out		
2	0	0
Check-Out		
3	0	0
Check-Out		
4	0	1
Check-Out		

	reservation_status_date
0	01-07-2015
1	01-07-2015
2	02-07-2015
3	02-07-2015
4	03-07-2015

[5 rows x 32 columns]

train.columns

```
Index(['hotel', 'is_canceled', 'lead_time', 'arrival_date_year',
      'arrival_date_month', 'arrival_date_week_number',
      'arrival_date_day_of_month', 'stays_in_weekend_nights',
      'stays_in_week_nights', 'adults', 'children', 'babies', 'meal',
      'country', 'market_segment', 'distribution_channel',
      'is_repeated_guest', 'previous_cancellations',
      'previous_bookings_not_canceled', 'reserved_room_type',
      'assigned_room_type', 'booking_changes', 'deposit_type',
      'agent',
      'company', 'days_in_waiting_list', 'customer_type', 'adr',
      'required_car_parking_spaces', 'total_of_special_requests',
      'reservation_status', 'reservation_status_date'],
      dtype='object')
```

train.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 10000 entries, 0 to 9999

Data columns (total 32 columns):

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	hotel	10000 non-null	object
1	is_canceled	10000 non-null	int64
2	lead_time	10000 non-null	int64
3	arrival_date_year	10000 non-null	int64
4	arrival_date_month	10000 non-null	object
5	arrival_date_week_number	10000 non-null	int64
6	arrival_date_day_of_month	10000 non-null	int64

7	stays_in_weekend_nights	10000	non-null	int64
8	stays_in_week_nights	10000	non-null	int64
9	adults	10000	non-null	int64
10	children	10000	non-null	int64
11	babies	10000	non-null	int64
12	meal	10000	non-null	object
13	country	9992	non-null	object
14	market_segment	10000	non-null	object
15	distribution_channel	10000	non-null	object
16	is_repeated_guest	10000	non-null	int64
17	previous_cancellations	10000	non-null	int64
18	previous_bookings_not_canceled	10000	non-null	int64
19	reserved_room_type	10000	non-null	object
20	assigned_room_type	10000	non-null	object
21	booking_changes	10000	non-null	int64
22	deposit_type	10000	non-null	object
23	agent	8471	non-null	float64
24	company	555	non-null	float64
25	days_in_waiting_list	10000	non-null	int64
26	customer_type	10000	non-null	object
27	adr	10000	non-null	float64
28	required_car_parking_spaces	10000	non-null	int64
29	total_of_special_requests	10000	non-null	int64
30	reservation_status	10000	non-null	object
31	reservation_status_date	10000	non-null	object

dtypes: float64(3), int64(17), object(12)

memory usage: 2.4+ MB

```
total = train.isnull().sum().sort_values(ascending=False)
```

```
train.replace("NULL",np.nan,inplace=True)
```

```
train.head(500)
```

	hotel	is_canceled	lead_time	arrival_date_year	\
0	Resort Hotel	0	342	2015	
1	Resort Hotel	0	737	2015	
2	Resort Hotel	0	7	2015	
3	Resort Hotel	0	13	2015	
4	Resort Hotel	0	14	2015	
...	
495	Resort Hotel	0	92	2015	
496	Resort Hotel	0	91	2015	
497	Resort Hotel	0	38	2015	
498	Resort Hotel	0	6	2015	
499	Resort Hotel	0	38	2015	

	arrival_date_month	arrival_date_week_number	arrival_date_day_of_month	\
0	July			27
1				

1	July	27
1		
2	July	27
1		
3	July	27
1		
4	July	27
1		
..
...		
495	July	29
17		
496	July	29
17		
497	July	29
17		
498	July	29
17		
499	July	29
17		

	stays_in_weekend_nights	stays_in_week_nights	adults	...	
deposit_type \					
0	0	0	2	...	No
Deposit					
1	0	0	2	...	No
Deposit					
2	0	1	1	...	No
Deposit					
3	0	1	1	...	No
Deposit					
4	0	2	2	...	No
Deposit					
..	
...					
495	1	2	2	...	No
Deposit					
496	1	2	2	...	No
Deposit					
497	1	2	2	...	No
Deposit					
498	1	2	2	...	No
Deposit					
499	1	2	2	...	No
Deposit					

	agent	company	days_in_waiting_list	customer_type	adr	\
0	NaN	NaN	0	Transient	0.00	
1	NaN	NaN	0	Transient	0.00	

2	NaN	NaN	0	Transient	75.00
3	304.0	NaN	0	Transient	75.00
4	240.0	NaN	0	Transient	98.00
..
495	240.0	NaN	0	Transient	103.50
496	240.0	NaN	0	Transient	118.50
497	240.0	NaN	0	Transient	192.00
498	142.0	NaN	0	Contract	128.27
499	240.0	NaN	0	Transient	135.00

	required_car_parking_spaces	total_of_special_requests	\
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	1
..
495	1	3	3
496	0	3	3
497	0	1	1
498	0	0	0
499	0	0	0

	reservation_status	reservation_status_date
0	Check-Out	01-07-2015
1	Check-Out	01-07-2015
2	Check-Out	02-07-2015
3	Check-Out	02-07-2015
4	Check-Out	03-07-2015
..
495	Check-Out	20-07-2015
496	Check-Out	20-07-2015
497	Check-Out	20-07-2015
498	Check-Out	20-07-2015
499	Check-Out	20-07-2015

[500 rows x 32 columns]

train.isnull()

	hotel	is_canceled	lead_time	arrival_date_year
arrival_date_month	\			
0	False	False	False	False
False				
1	False	False	False	False
False				
2	False	False	False	False
False				
3	False	False	False	False
False				

4	False	False	False	False
False				
...
...				
9995	False	False	False	False
False				
9996	False	False	False	False
False				
9997	False	False	False	False
False				
9998	False	False	False	False
False				
9999	False	False	False	False
False				

	arrival_date_week_number	arrival_date_day_of_month	\
0	False	False	
1	False	False	
2	False	False	
3	False	False	
4	False	False	
...	
9995	False	False	
9996	False	False	
9997	False	False	
9998	False	False	
9999	False	False	

	stays_in_weekend_nights	stays_in_week_nights	adults	...	\
0	False	False	False	...	
1	False	False	False	...	
2	False	False	False	...	
3	False	False	False	...	
4	False	False	False	...	
...	
9995	False	False	False	...	
9996	False	False	False	...	
9997	False	False	False	...	
9998	False	False	False	...	
9999	False	False	False	...	

	deposit_type	agent	company	days_in_waiting_list
customer_type	\			
0	False	True	True	False
False				
1	False	True	True	False
False				
2	False	True	True	False
False				
3	False	False	True	False

False					
4	False	False	True	False	
False					
...
.					
9995	False	False	True	False	
False					
9996	False	False	True	False	
False					
9997	False	True	True	False	
False					
9998	False	False	True	False	
False					
9999	False	False	True	False	
False					

	adr	required_car_parking_spaces	total_of_special_requests	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	
...	
9995	False	False	False	
9996	False	False	False	
9997	False	False	False	
9998	False	False	False	
9999	False	False	False	

	reservation_status	reservation_status_date
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False
...
9995	False	False
9996	False	False
9997	False	False
9998	False	False
9999	False	False

[10000 rows x 32 columns]

train.notnull()

	hotel	is_canceled	lead_time	arrival_date_year
arrival_date_month				\
0	True	True	True	True
True				

1	True	True	True	True
True				
2	True	True	True	True
True				
3	True	True	True	True
True				
4	True	True	True	True
True				
...
...				
9995	True	True	True	True
True				
9996	True	True	True	True
True				
9997	True	True	True	True
True				
9998	True	True	True	True
True				
9999	True	True	True	True
True				
arrival_date_week_number arrival_date_day_of_month \				
0		True	True	
1		True	True	
2		True	True	
3		True	True	
4		True	True	
...		
9995		True	True	
9996		True	True	
9997		True	True	
9998		True	True	
9999		True	True	
stays_in_weekend_nights stays_in_week_nights adults ... \				
0		True	True	True
1		True	True	True
2		True	True	True
3		True	True	True
4		True	True	True
...	
9995		True	True	True
9996		True	True	True
9997		True	True	True
9998		True	True	True
9999		True	True	True
deposit_type agent company days_in_waiting_list				
customer_type	adr	\		
0	True	False	False	True

True	True					
1		True	False	False		True
True	True					
2		True	False	False		True
True	True					
3		True	True	False		True
True	True					
4		True	True	False		True
True	True					
...	
.	...					
9995		True	True	False		True
True	True					
9996		True	True	False		True
True	True					
9997		True	False	False		True
True	True					
9998		True	True	False		True
True	True					
9999		True	True	False		True
True	True					

	required_car_parking_spaces	total_of_special_requests	\
0	True	True	
1	True	True	
2	True	True	
3	True	True	
4	True	True	
...	
9995	True	True	
9996	True	True	
9997	True	True	
9998	True	True	
9999	True	True	

	reservation_status	reservation_status_date
0	True	True
1	True	True
2	True	True
3	True	True
4	True	True
...
9995	True	True
9996	True	True
9997	True	True
9998	True	True
9999	True	True

[10000 rows x 32 columns]

```
miss=train.isnull()
```

```
avg=train["agent"].astype("float").mean(axis=0)
```

```
avg1=train["company"].astype("float").mean(axis=0)
```

```
avg1
```

```
234.38558558558557
```

```
train['agent'].replace(np.nan,avg,inplace=True)
```

```
train['company'].replace(np.nan,avg1,inplace=True)
```

```
train.head(100)
```

	hotel	is_canceled	lead_time	arrival_date_year	\
0	Resort Hotel	0	342	2015	
1	Resort Hotel	0	737	2015	
2	Resort Hotel	0	7	2015	
3	Resort Hotel	0	13	2015	
4	Resort Hotel	0	14	2015	
..	
95	Resort Hotel	1	72	2015	
96	Resort Hotel	1	63	2015	
97	Resort Hotel	0	63	2015	
98	Resort Hotel	0	101	2015	
99	Resort Hotel	0	102	2015	

	arrival_date_month	arrival_date_week_number
arrival_date_day_of_month	\	
0	July	27
1		
1	July	27
1		
2	July	27
1		
3	July	27
1		
4	July	27
1		
..
...		
95	July	27
3		
96	July	27
3		
97	July	27
3		
98	July	27
3		
99	July	27
3		

	stays_in_weekend_nights	stays_in_week_nights	adults	...	
deposit_type \					
0	0	0	2	...	No
Deposit					
1	0	0	2	...	No
Deposit					
2	0	1	1	...	No
Deposit					
3	0	1	1	...	No
Deposit					
4	0	2	2	...	No
Deposit					
..	
...					
95	0	2	2	...	No
Deposit					
96	2	5	2	...	No
Deposit					
97	2	5	3	...	No
Deposit					
98	2	5	2	...	No
Deposit					
99	2	5	2	...	No
Deposit					
	agent	company	days_in_waiting_list	customer_type	adr
\					
0	214.417896	234.385586	0	Transient	0.00
1	214.417896	234.385586	0	Transient	0.00
2	214.417896	234.385586	0	Transient	75.00
3	304.000000	234.385586	0	Transient	75.00
4	240.000000	234.385586	0	Transient	98.00
..
95	240.000000	234.385586	0	Transient	73.80
96	242.000000	234.385586	0	Transient	117.00
97	105.000000	234.385586	0	Transient	196.54
98	240.000000	234.385586	0	Transient	99.30
99	250.000000	234.385586	0	Transient	90.95

	required_car_parking_spaces	total_of_special_requests	\
0	0	0	
1	0	0	
2	0	0	
3	0	0	
4	0	1	
..	
95	0	1	
96	0	1	
97	0	1	
98	1	2	
99	0	0	

	reservation_status	reservation_status_date
0	Check-Out	01-07-2015
1	Check-Out	01-07-2015
2	Check-Out	02-07-2015
3	Check-Out	02-07-2015
4	Check-Out	03-07-2015
..
95	Canceled	29-06-2015
96	Canceled	13-05-2015
97	Check-Out	10-07-2015
98	Check-Out	10-07-2015
99	Check-Out	10-07-2015

[100 rows x 32 columns]

#Data Integration

```
customers_data = {
    'CustomerID': [101, 102, 103, 104, 105],
    'Name': ['Alice', 'Bob', 'Charlie', 'David', 'Eve'],
    'Age': [25, 32, 50, 45, 28],
    'Email': ['alice@example.com', 'bob@example.com',
'charlie@example.com', 'david@example.com', 'eve@example.com'],
    'Gender': ['Female', 'Male', 'Male', 'Male', 'Female'],
    'Income': [50000, 75000, 60000, np.nan, 80000]
}
```

```
customers_df = pd.DataFrame(customers_data)
```

Create the product inventory dataset

```
products_data = {
    'ProductID': [501, 502, 503, 504, 505],
    'Product_Name': ['Laptop', 'Phone', 'Tablet', 'Keyboard',
'Mouse'],
    'Category': ['Electronics', 'Electronics', 'Electronics',
'Accessories', 'Accessories'],
    'Price': [1200, 800, 400, 50, 20],
}
```

```

    'Stock_Quantity': [15, 20, 10, 100, 50]
}

```

```

products_df = pd.DataFrame(products_data)

```

```

sales_data = {
    'SaleID': [1, 2, 3, 4, 5],
    'CustomerID': [101, 102, 103, 110, 105],
    'ProductID': [502, 504, 503, 502, 505],
    'Quantity': [2, 1, 3, 2, 5]
}

```

```

sales_df = pd.DataFrame(sales_data)

```

```

final_df=sales_df.merge(customers_df,on='CustomerID')
final_df=final_df.merge(products_df,on='ProductID')

```

```

final_df

```

	SaleID	CustomerID	ProductID	Quantity	Name	Age
Email \						
0	1	101	502	2	Alice	25
alice@example.com						
1	2	102	504	1	Bob	32
bob@example.com						
2	3	103	503	3	Charlie	50
charlie@example.com						
3	5	105	505	5	Eve	28
eve@example.com						

	Gender	Income	Product_Name	Category	Price	Stock_Quantity
0	Female	50000.0	Phone	Electronics	800	20
1	Male	75000.0	Keyboard	Accessories	50	100
2	Male	60000.0	Tablet	Electronics	400	10
3	Female	80000.0	Mouse	Accessories	20	50

```

final_df['Sales']=final_df['Price']*final_df['Quantity']

```

```

final_df

```

	SaleID	CustomerID	ProductID	Quantity	Name	Age
Email \						
0	1	101	502	2	Alice	25
alice@example.com						
1	2	102	504	1	Bob	32
bob@example.com						
2	3	103	503	3	Charlie	50
charlie@example.com						
3	5	105	505	5	Eve	28
eve@example.com						

	Gender	Income	Product_Name	Category	Price	Stock_Quantity
0	Female	50000.0	Phone	Electronics	800	20
1	Male	75000.0	Keyboard	Accessories	50	100
2	Male	60000.0	Tablet	Electronics	400	10
3	Female	80000.0	Mouse	Accessories	20	50

#Data Transformation

```
final_df.columns
```

```
Index(['SaleID', 'CustomerID', 'ProductID', 'Quantity', 'Name', 'Age', 'Email', 'Gender', 'Income', 'Product_Name', 'Category', 'Price', 'Stock_Quantity', 'Sales'], dtype='object')
```

```
final_df=final_df.groupby(['CustomerID', 'ProductID', 'Name', 'Age', 'Email', 'Gender', 'Income', 'Category']).agg({'Quantity': 'sum', 'Sales': 'sum', 'Stock_Quantity': 'sum'}).reset_index()
```

```
final_df
```

	CustomerID	ProductID	Name	Age	Email	Gender
0	101	502	Alice	25	alice@example.com	Female
1	102	504	Bob	32	bob@example.com	Male
2	103	503	Charlie	50	charlie@example.com	Male
3	105	505	Eve	28	eve@example.com	Female

	Category	Quantity	Sales	Stock_Quantity
0	Electronics	2	1600	20
1	Accessories	1	50	100
2	Electronics	3	1200	10
3	Accessories	5	100	50

#Data Discretization

```
df=pd.read_csv('IBRD.csv', nrows=60)
df.head(60)
```

	Category	Code	Category	Subcategory	Fiscal Year	\
0		a	Assets	Due from Banks	2010	
1		a	Assets	Due from Banks	2009	
2		a	Assets	Due from Banks	2010	
3		a	Assets	Due from Banks	2009	
4		a	Assets	Investments	2010	
5		a	Assets	Investments	2009	
6		a	Assets	Securities	2010	
7		a	Assets	Securities	2009	
8		a	Assets	Nonnegotiable	2010	
9		a	Assets	Nonnegotiable	2009	
10		a	Assets	Derivative Assets	2010	
11		a	Assets	Derivative Assets	2009	
12		a	Assets	Derivative Assets	2010	
13		a	Assets	Derivative Assets	2009	
14		a	Assets	Derivative Assets	2010	
15		a	Assets	Derivative Assets	2009	
16		a	Assets	Derivative Assets	2010	
17		a	Assets	Derivative Assets	2009	
18		a	Assets	Receivables	2010	
19		a	Assets	Receivables	2009	
20		a	Assets	Other Receivables	2010	
21		a	Assets	Other Receivables	2009	
22		a	Assets	Other Receivables	2010	
23		a	Assets	Other Receivables	2009	
24		a	Assets	Loans Outstanding	2010	
25		a	Assets	Loans Outstanding	2009	
26		a	Assets	Other Assets	2010	
27		a	Assets	Other Assets	2009	
28		a	Assets	Other Assets	2010	
29		a	Assets	Other Assets	2009	
30		a	Assets	Other Assets	2010	
31		a	Assets	Other Assets	2009	
32		l	Liabilities	Borrowings	2010	
33		l	Liabilities	Borrowings	2009	
34		l	Liabilities	Sold or Lent	2010	
35		l	Liabilities	Sold or Lent	2009	
36		l	Liabilities	Derivative Liabilities	2010	
37		l	Liabilities	Derivative Liabilities	2009	
38		l	Liabilities	Derivative Liabilities	2010	
39		l	Liabilities	Derivative Liabilities	2009	
40		l	Liabilities	Derivative Liabilities	2010	
41		l	Liabilities	Derivative Liabilities	2009	
42		l	Liabilities	Derivative Liabilities	2010	
43		l	Liabilities	Derivative Liabilities	2009	
44		l	Liabilities	Other	2010	
45		l	Liabilities	Other Liabilities	2010	
46		l	Liabilities	Other Liabilities	2009	
47		l	Liabilities	Other Liabilities	2010	
48		l	Liabilities	Other Liabilities	2009	

49	l	Liabilities	Other Liabilities	2010
50	l	Liabilities	Other Liabilities	2009
51	l	Liabilities	Other Liabilities	2010
52	l	Liabilities	Other Liabilities	2009
53	e	Equity	Capital Stock	2010
54	e	Equity	Capital Stock	2009
55	e	Equity	Deferred Amounts	2010
56	e	Equity	Deferred Amounts	2009
57	e	Equity	Retained Earnings	2010
58	e	Equity	Retained Earnings	2009
Amount (US\$, Millions)				
0		1581		
1		2380		
2		222		
3		664		
4		36012		
5		41012		
6		289		
7		33		
8		1123		
9		1202		
10		13249		
11		18467		
12		17633		
13		19559		
14		87457		
15		82793		
16		3287		
17		2246		
18		171		
19		176		
20		47		
21		95		
22		764		
23		889		
24		118104		
25		103657		
26		0		
27		325		
28		635		
29		625		
30		2436		
31		1297		
32		128577		
33		110040		
34		998		
35		2323		
36		13360		
37		18923		

38	17623
39	19551
40	78655
41	76321
42	780
43	847
44	8
45	307
46	2457
47	1190
48	1495
49	1164
50	662
51	2793
52	2707
53	11492
54	11491
55	313
56	359
57	28793
58	29870

```
pd.qcut(df["Amount (US$, Millions)"],q=4).head()
```

```
0    (648.5, 2246.0]
1    (2246.0, 18695.0]
2    (-0.001, 648.5]
3    (648.5, 2246.0]
4    (18695.0, 128577.0]
```

```
Name: Amount (US$, Millions), dtype: category
```

```
Categories (4, interval[float64, right]): [(-0.001, 648.5] < (648.5,
2246.0] < (2246.0, 18695.0] <
(18695.0, 128577.0]]
```

```
df["Amount (US$, Millions)"]=pd.cut(df["Amount (US$,
Millions)"],bins=[0,648.5,2246,18695,128577],labels=["low","Medium","H
igh","Very High"])
print(df.head(60))
```

	Category	Code	Category	Subcategory	Fiscal Year	\
0		a	Assets	Due from Banks	2010	
1		a	Assets	Due from Banks	2009	
2		a	Assets	Due from Banks	2010	
3		a	Assets	Due from Banks	2009	
4		a	Assets	Investments	2010	
5		a	Assets	Investments	2009	
6		a	Assets	Securities	2010	
7		a	Assets	Securities	2009	
8		a	Assets	Nonnegotiable	2010	
9		a	Assets	Nonnegotiable	2009	

10	a	Assets	Derivative Assets	2010
11	a	Assets	Derivative Assets	2009
12	a	Assets	Derivative Assets	2010
13	a	Assets	Derivative Assets	2009
14	a	Assets	Derivative Assets	2010
15	a	Assets	Derivative Assets	2009
16	a	Assets	Derivative Assets	2010
17	a	Assets	Derivative Assets	2009
18	a	Assets	Receivables	2010
19	a	Assets	Receivables	2009
20	a	Assets	Other Receivables	2010
21	a	Assets	Other Receivables	2009
22	a	Assets	Other Receivables	2010
23	a	Assets	Other Receivables	2009
24	a	Assets	Loans Outstanding	2010
25	a	Assets	Loans Outstanding	2009
26	a	Assets	Other Assets	2010
27	a	Assets	Other Assets	2009
28	a	Assets	Other Assets	2010
29	a	Assets	Other Assets	2009
30	a	Assets	Other Assets	2010
31	a	Assets	Other Assets	2009
32	l	Liabilities	Borrowings	2010
33	l	Liabilities	Borrowings	2009
34	l	Liabilities	Sold or Lent	2010
35	l	Liabilities	Sold or Lent	2009
36	l	Liabilities	Derivative Liabilities	2010
37	l	Liabilities	Derivative Liabilities	2009
38	l	Liabilities	Derivative Liabilities	2010
39	l	Liabilities	Derivative Liabilities	2009
40	l	Liabilities	Derivative Liabilities	2010
41	l	Liabilities	Derivative Liabilities	2009
42	l	Liabilities	Derivative Liabilities	2010
43	l	Liabilities	Derivative Liabilities	2009
44	l	Liabilities	Other	2010
45	l	Liabilities	Other Liabilities	2010
46	l	Liabilities	Other Liabilities	2009
47	l	Liabilities	Other Liabilities	2010
48	l	Liabilities	Other Liabilities	2009
49	l	Liabilities	Other Liabilities	2010
50	l	Liabilities	Other Liabilities	2009
51	l	Liabilities	Other Liabilities	2010
52	l	Liabilities	Other Liabilities	2009
53	e	Equity	Capital Stock	2010
54	e	Equity	Capital Stock	2009
55	e	Equity	Deferred Amounts	2010
56	e	Equity	Deferred Amounts	2009
57	e	Equity	Retained Earnings	2010
58	e	Equity	Retained Earnings	2009

Amount (US\$, Millions)	
0	Medium
1	High
2	low
3	Medium
4	Very High
5	Very High
6	low
7	low
8	Medium
9	Medium
10	High
11	High
12	High
13	Very High
14	Very High
15	Very High
16	High
17	Medium
18	low
19	low
20	low
21	low
22	Medium
23	Medium
24	Very High
25	Very High
26	NaN
27	low
28	low
29	low
30	High
31	Medium
32	Very High
33	Very High
34	Medium
35	High
36	High
37	Very High
38	High
39	Very High
40	Very High
41	Very High
42	Medium
43	Medium
44	low
45	low
46	High

47	Medium
48	Medium
49	Medium
50	Medium
51	High
52	High
53	High
54	High
55	low
56	low
57	Very High
58	Very High