Perspectives on Data

Tradeoffs and considerations in data modeling





Types of Data

Tabular Data: Tables, Spreadsheets, DataFrames

Semi-Structured Data: JSON, XML

Unstructured Data: Text, Video, Images, Raw Sensor Data



Tabular Data

A rectangular data structure where rows correspond to observations and each column corresponds to properties (attributes) of the observation.

	Customers							
4		CustomerId -	FirstName +	LastName -	DateCreated -	Cli		
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM			
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM			
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM			
	+	4	Brian	Griffin	13/06/2014 9:16:36 PM			
	+	5	Cosmo	Kramer	13/06/2014 9:16:41 PM			
	+	6	Philip	Fry	13/06/2014 9:17:02 PM			
	+	7	Amy	Wong	13/06/2014 9:22:05 PM			
	+	8	Hubert J.	Farnsworth	13/06/2014 9:22:19 PM			
	+	9	Marge	Simpson	13/06/2014 9:22:37 PM			
	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM			
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM			
*		(New)			15/06/2014 9:00:01 PM			



Tabular Data

Characteristic I. Named Attributes

- The columns names define all of the relevant attributes for each observation.
- Values are accessed by name.
- Every row has the same set of attributes (if data are missing, value is explicitly NULL)

	Customers							
4		CustomerId -	FirstName 🕶	LastName →	DateCreated →	Cli		
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM			
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM			



Tabular Data

Characteristic 2. Atomic Values

- Each cell (a row, column pair) is generally considered to be an atomic value---i.e., it is not readily divisible--such as, an integer, a date, a string.
- Each column has a defined data type consistent across all rows.

	Customers						
4		CustomerId -	FirstName +	LastName +	DateCreated -	Cli	
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM		
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM		
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM		
	+	4	Brian	Griffin	13/06/2014 9:16:36 PM		
	+	5	Cosmo	Kramer	13/06/2014 9:16:41 PM		
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	+	9	Marge	Simpson	13/06/2014 9:22:37 PM		
	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM		
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM		
*		(New)			15/06/2014 9:00:01 PM		



Schema

An abstract definition of a tabular dataset consisting of a set of named attributes and their corresponding data types.

Car(Make: String, Model: String, Year: Int)

Person(First: String, Last: String, SSN: String)

Grade(SID: String, Class: String, Score: Float)



Data Model

Key advantage is that you can separate how users program against the data from how the data are actually stored.

Logical data model: describes the semantics of the data, as represented by the particular data analysis technology.

Physical data model: Describes the physical means by which data are stored in terms of bits and organization.



Semi-Structured Data

A data structure where each row has a variable set of attributes the values are possibly not atomic.

```
{'name': 'Car1', 'features': ['BackupCamera', 'CruiseControl']}
{'name': 'Car2', 'features': ['BackupCamera', 'ParkingAssist',
'CruiseControl']}
```

```
- <Parts>
 - <Part>
     <Id>4478</Id>
     <Part_Name>1000 Ohm Resistor</Part_Name>
     <Total_Available>25000</Total_Available>
     <Price>0.01</Price>
   </Part>
 - <Part>
     <Id>3328</Id>
     <Part_Name>15000 Ohm Resistor</Part_Name>
     <Total_Available>75000</Total_Available>
     <Price>0.02</Price>
   </Part>
 - <Part>
     <Id>4725</Id>
     <Part_Name>555 Timer IC</Part_Name>
     <Total_Available>1500</Total_Available>
     <Price>0.25</Price>
   </Part>
  </Parts>
```

Semi-Structured Data

Characteristic I. Named Attributes

- Attributes are still named!
- But each row may have a different set of named attributes.

```
- <Parts>
 - <Part>
     <Id>4478</Id>
     <Part_Name>1000 Ohm Resistor
     <Total Available>25000</Total Available>
     <Price>0.01</Price>
   </Part>
 - <Part>
     <Id>3328</Id>
     <Part_Name>15000 Ohm Resistor</Part_Name>
     <Total_Available>75000</Total_Available>
     <Price>0.02</Price>
   </Part>
 - <Part>
     <Id>4725</Id>
     <Part_Name>555 Timer IC</Part_Name>
     <Total_Available>1500</Total_Available>
     <Price>0.25</Price>
   </Part>
 </Parts>
```



Semi-Structured Data

Characteristic 2. Nested Values

- Values are not atomic and can be nested data structures.
- Data type is generally inferred on the fly.

```
- <Parts>
 - <Part>
     <Id>4478</Id>
     <Part_Name>1000 Ohm Resistor</Part_Name>
     <Total Available>25000</Total Available>
     <Price>0.01</Price>
   </Part>
 - <Part>
     <Id>3328</Id>
     <Part_Name>15000 Ohm Resistor</Part_Name>
     <Total_Available>75000</Total_Available>
     <Price>0.02</Price>
   </Part>
 - <Part>
     <Id>4725</Id>
     <Part_Name>555 Timer IC</Part_Name>
     <Total_Available>1500</Total_Available>
     <Price>0.25</Price>
   </Part>
 </Parts>
```



Semi-Structured Schema

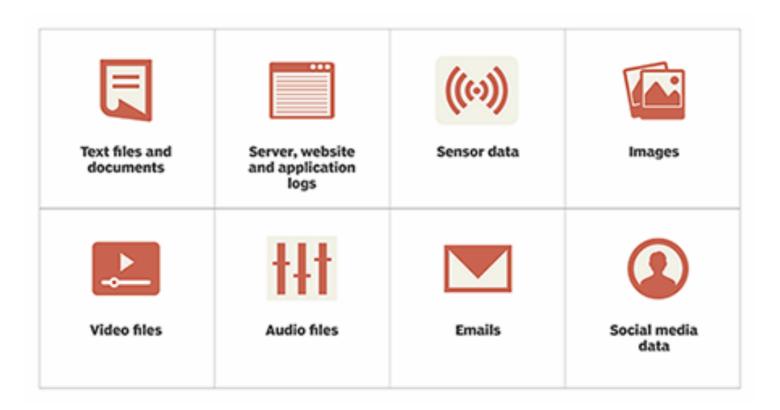
Can think of each row as defining its own schema and that definition is stored with each row.

```
- <Parts>
 - <Part>
     <Id>4478</Id>
     <Part_Name>1000 Ohm Resistor</Part_Name>
     <Total_Available>25000</Total_Available>
     <Price>0.01</Price>
   </Part>
 - <Part>
     <Id>3328</Id>
     <Part_Name>15000 Ohm Resistor</Part_Name>
     <Total_Available>75000</Total_Available>
     <Price>0.02</Price>
   </Part>
 - <Part>
     <Id>4725</Id>
     <Part_Name>555 Timer IC</Part_Name>
     <Total_Available>1500</Total_Available>
     <Price>0.25</Price>
   </Part>
 </Parts>
```



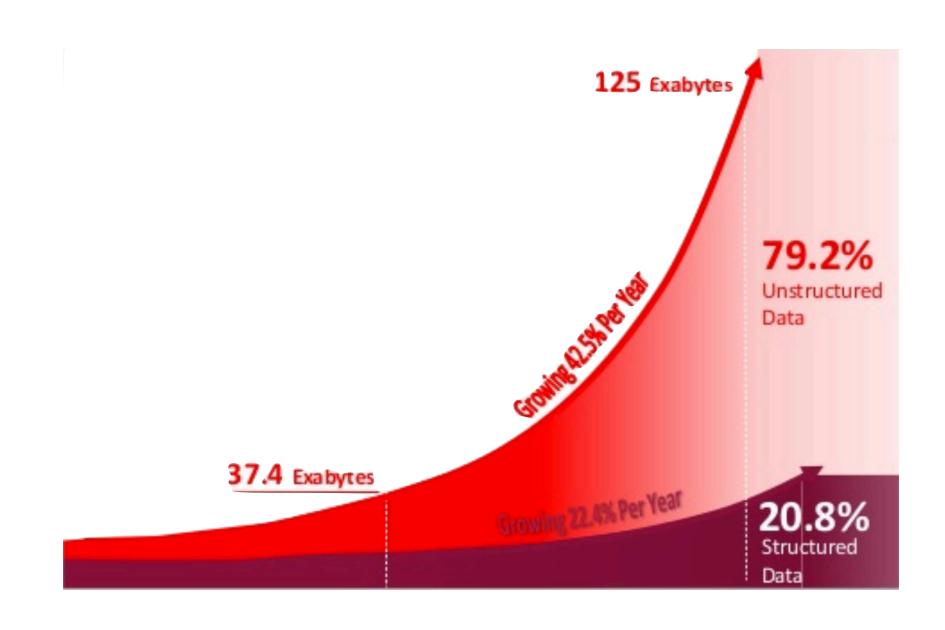
Unstructured Data

Everything else!





Unstructured Data





Tradeoffs

Most organizations have a mix of different types of data!

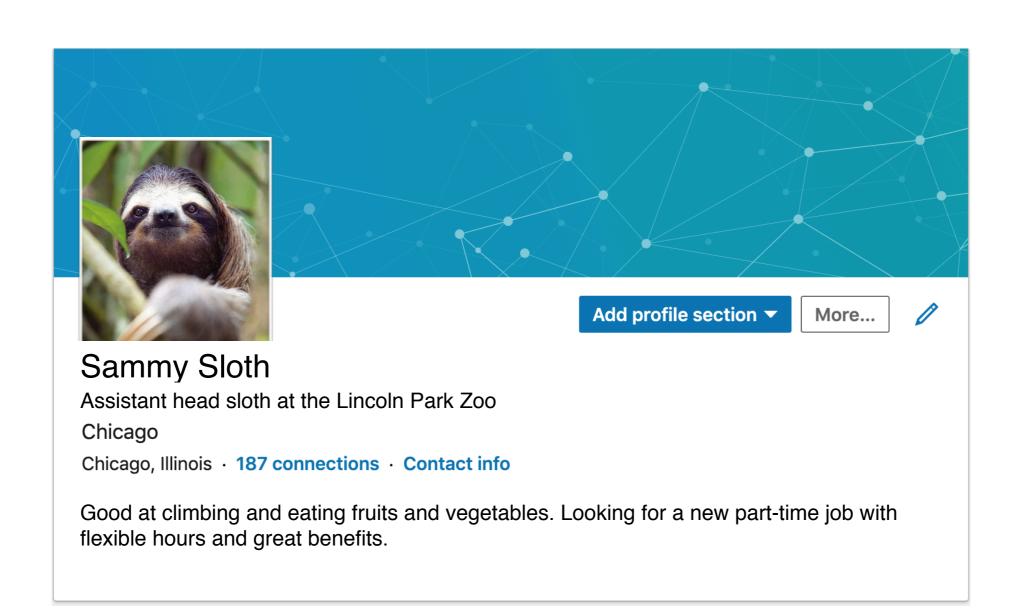
Tabular Data (structured): Tables, Spreadsheets, DataFrames

- + Simplified programming model (Pandas, SQL, Excel)
- + Efficient and scalable
- Inflexible (need to define a schema up front)
- Hard to migrate data into other systems

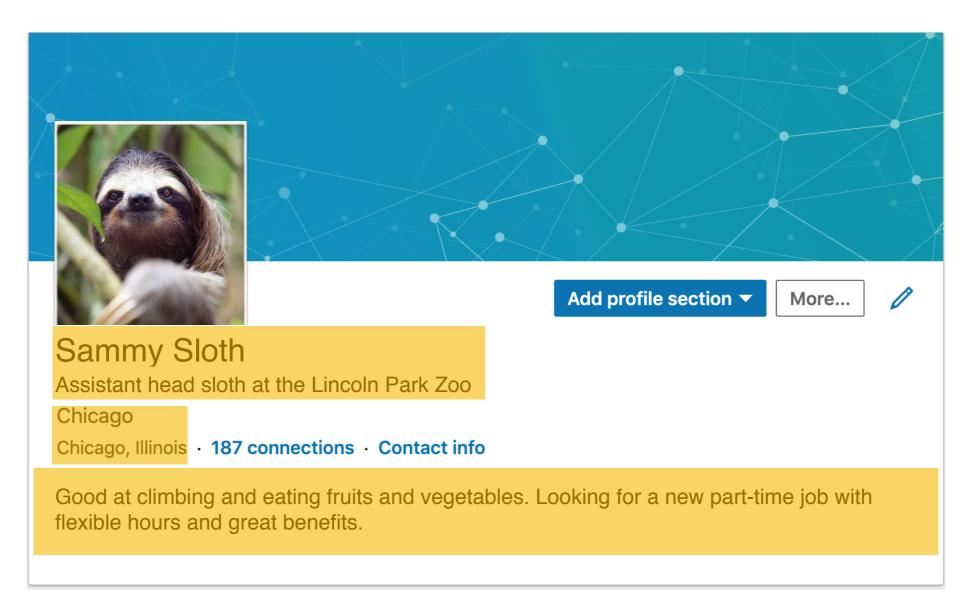
Semi-Structured Data: JSON, XML

- + Very flexible (on-the-fly schema design)
- + Portable (readable by all major programming languages)
- Hard to debug code with complex failure modes
- Extra data and extra code









Semi-Structured Data



```
{'name': 'Sammy Sloth',
  'location_city': 'Chicago',
  'location_state': 'Illinois',
  'position': 'Assistant head sloth at the Lincoln Park Zoo',
  'looking_for': 'Good at climbing and eating fruits and
  vegetables. Looking for a new part-time job with flexible
  hours and great benefits.'
}
```

- + Flexible
- + Redesign of the website doesn't need a data change
- + Portable



Count the number of climbers per city

```
cities = {}

for a in Animals:
   obj = json.loads(a)

  city = obj['location_city']
   is_climber = obj['looking_for'].contains('climber')

  if city not in cities:
      cities[city] = 0

  cities[city] += (is_climber)
```



Count the number of climbers per **state**

```
states = {}

for a in Animals:
   obj = json.loads(a)

  state = obj['location_state']
   is_climber = obj['looking_for'].contains('climber')

  if state not in states:
      states[state] = 0

  states[state] += (is_climber)
```



Extracting structured tables from semi-structured or unstructured data simplifies analysis later on

```
{'name': 'Sammy Sloth',
  'location_city': 'Chicago',
  'location_state': 'Illinois',
  'position': 'Assistant head sloth at the Lincoln Park Zoo',
  'looking_for': 'Good at climbing and eating fruits and
  vegetables. Looking for a new part-time job with flexible hours
  and great benefits.'
}
```

Talent(Name: String, City: String, State: String, Climbing: Boolean)



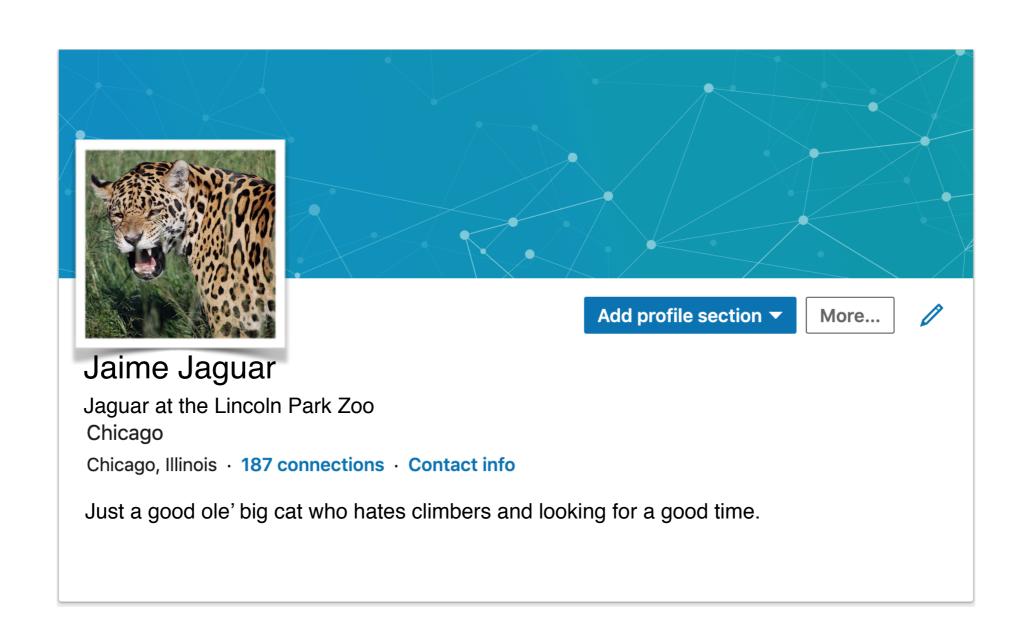
Count the number of climbers per city

Talent(Name: String, City: String, State: String, Climbing: Boolean)

```
df.group_by('city').sum('climbing')
df.group_by('state').sum('climbing')
```

+ Cost of structuring data amortizes over all the analyses you want to do







Format conversions often lose or misrepresent data

```
'looking_for': 'Just a good ole' big cat who hates climbers and looking for a good time.'
```

Talent(Name: String, City: String, State: String, Climbing: Boolean)



Paradox of Structured Data

Always certain, seldom right

- Query languages (like SQL) are built on the principles of formal logic
- A database holds facts and allows users to make principled inferences from these facts



Paradox of Structured Data

- Mary is a citizen of France
- Jenny is a citizen of France
- •All French citizens wear hats

People

Name	Citizenship		
Mary	France		
Jenny	France		

Countries

Citizenship	Wears	
France	hats	

What is Mary's citizenship?

SELECT Citizenship
FROM People
WHERE Name='Mary';

people[people['name']='Mary']

Do both Mary and Jenny wear hats?

people.merge(countries)

Paradox of Structured Data

```
Does Peter Wear A Hat? 

SELECT Name, Wears

FROM People, Countries

WHERE People.Citizenship =

Countries.Citizenship AND

People = 'Peter';

Peter = people[people['name']='Peter']

peter = people[people['name']='Peter']
```

Closed World Assumption: Everything not currently known to be true, is false



Up-to the developer to understand and account for uncertainty

```
'looking_for': 'Just a good ole' big cat who hates climbers and looking for a good time.'
```

Talent(Name: String, City: String, State: String, Climbing: Boolean)



 Schemas not only should include attribute types but also constraints on the values they can take.

Person(First: String, Last: String, SSN: String)



 Schemas not only should include attribute types but also constraints on the values they can take.

Person(First: String, Last: String, SSN: String)

Syntactic Constraints

- First name: must be alphabetical (upto apostrophes and dashes)
- Last name: must be alphabetical (upto apostrophes and dashes)
- SSN: must follow a format XXX-XX-XXXX

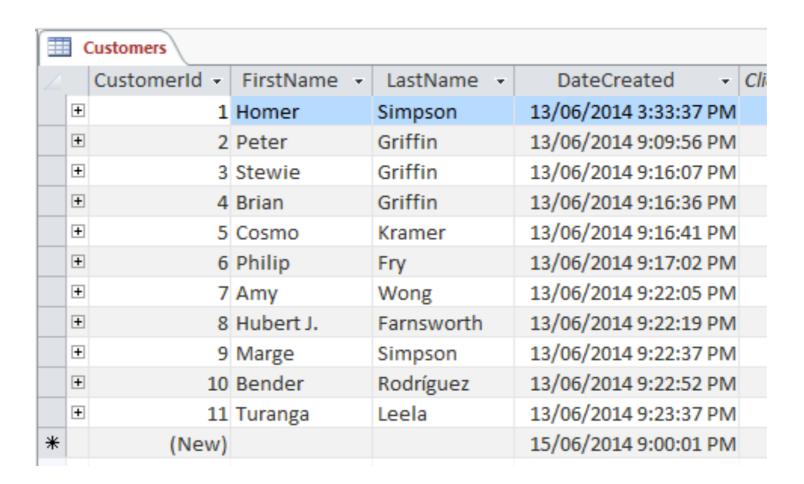
Semantic Constrains

SSNs must be unique



Syntactic Constraints: Formatting of a single cell of data

Semantic Constrains: Relationships between data





Syntactic Constraints: Formatting of a single cell of data

No cell can be empty

	Customers						
2		CustomerId -	FirstName 🔻	LastName -	DateCreated -	Cli	
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM		
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM		
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM		
	+	4	Brian	Griffin	13/06/2014 9:16:36 PM		
	+	5	Cosmo	Kramer	13/06/2014 9:16:41 PM		
	+	6	Philip	Fry	13/06/2014 9:17:02 PM		
	+	7	Amy	Wong	13/06/2014 9:22:05 PM		
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	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM		
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM		
*		(New)			15/06/2014 9:00:01 PM		



Semantic Constraints: Relationships between data

All firstname, lastname pairs are unique

	Tustomers Customers							
4		CustomerId -	FirstName -	LastName -	DateCreated -	Cli		
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM			
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM			
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM			
	+	4	Brian	Griffin	13/06/2014 9:16:36 PM			
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	+	7	Amy	Wong	13/06/2014 9:22:05 PM			
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	+	9	Marge	Simpson	13/06/2014 9:22:37 PM			
	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM			
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM			
*		(New)			15/06/2014 9:00:01 PM			



Tradeoffs

Integrity constraints are really only useful over tabular data!

Tabular Data (structured): Tables, Spreadsheets, DataFrames

- + Simplified programming model (Pandas, SQL, Excel)
- + Efficient and scalable
- + Verifiable/Testable
- Inflexible (need to define a schema up front)
- Hard to migrate data into other systems

Semi-Structured Data: JSON, XML

- + Very flexible (on-the-fly schema design)
- + Portable (readable by all major programming languages)
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Semantic Constraints: Relationships between data

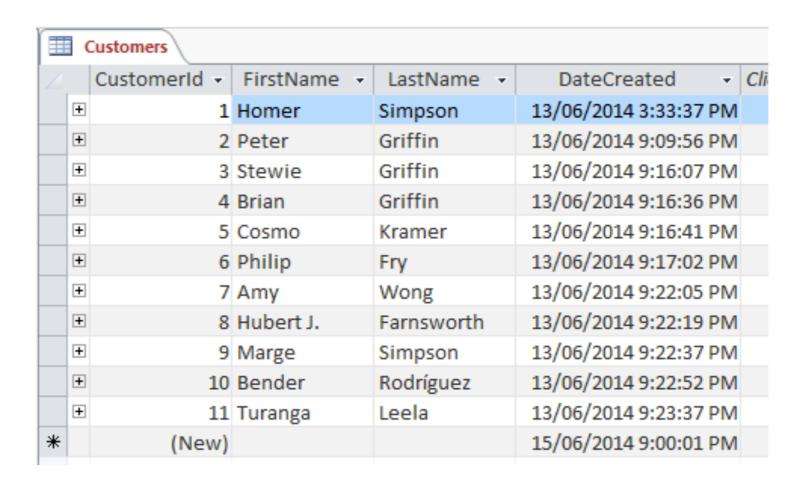
All firstname, lastname pairs are unique





Keys: A set of attributes such that no two records have the same values.

{CustomerID}, {FirstName, LastName} Is this it..?





Keys: A key + any other attribute is another key

{FirstName, LastName, DateCreated}

	Tustomers Customers						
4		CustomerId -	FirstName +	LastName -	DateCreated -	Cli	
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM		
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM		
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM		
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	+	11	Turanga	Leela	13/06/2014 9:23:37 PM		
*		(New)			15/06/2014 9:00:01 PM		



Candidate Key: A key that doesn't contain any other keys

{CustomerID}, {FirstName, LastName}, {DateCreated}

	Tustomers Customers						
4		CustomerId -	FirstName +	LastName -	DateCreated -	Cli	
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM		
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM		
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	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM		
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM		
*		(New)			15/06/2014 9:00:01 PM		



Any one of these column sets can uniquely "index" the records.

{CustomerID}, {FirstName, LastName}, {DateCreated}

	The Customers Customers					
4		CustomerId -	FirstName +	LastName +	DateCreated -	Cli
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM	
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM	
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM	
	+	4	Brian	Griffin	13/06/2014 9:16:36 PM	
	+	5	Cosmo	Kramer	13/06/2014 9:16:41 PM	
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	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM	
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM	
*		(New)			15/06/2014 9:00:01 PM	



Why might the creation date be a bad key? First/Last Name?

{CustomerID}, {FirstName, LastName}, {DateCreated}

	Customers						
4		CustomerId -	FirstName +	LastName -	DateCreated → Cl		
	+	1	Homer	Simpson	13/06/2014 3:33:37 PM		
	+	2	Peter	Griffin	13/06/2014 9:09:56 PM		
	+	3	Stewie	Griffin	13/06/2014 9:16:07 PM		
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	+	10	Bender	Rodríguez	13/06/2014 9:22:52 PM		
	+	11	Turanga	Leela	13/06/2014 9:23:37 PM		
*		(New)			15/06/2014 9:00:01 PM		



"Primary" Keys

 Schemas not only should include attribute types but also constraints on the values they can take.

Person(First: String, Last: String, SSN: String)

 Schemas should also describe what is the primary key for the table (uniquely identifies records)

Person(First: String, Last: String, SSN: String)