

HBase Schema Design





HBase Schema Design

How I Learned To Stop Worrying And Love Denormalization



What Is Schema Design?



Who am I?

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What Is Schema Design? Logical Data Modeling



What Is Schema Design? Logical Data Modeling + Physical Implementation



You always start with a logical model. Even if it's just an implicit one.

That's totally fine. (If you're right.)



There are lots of ways to model data. The most common one is:

Entity / Attribute / Relationship

(This is probably what you know of as just "data modeling".)



There's a well established visual language for data modeling.



Entities are boxes. With rounded corners if you're fancy.





Attributes are listed vertically in the box. Optionally with data types, for clarity.





Relationships are connecting lines. Optionally with special endpoints, and/or verbs













Hometown Genre







For relational databases, you usually start with this **normalized** model, then plug & chug.



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 $\begin{array}{l} \mbox{Entities} \rightarrow \mbox{Tables} \\ \mbox{Attributes} \rightarrow \mbox{Columns} \\ \mbox{Relationships} \rightarrow \mbox{Foreign Keys} \\ \mbox{Many-to-many} \rightarrow \mbox{Junction tables} \\ \mbox{Natural keys} \rightarrow \mbox{Artificial IDs} \end{array}$







So, what's missing?



So, what's missing? If your data is not massive, **NOTHING.**

You should use a relational database. They rock*



So, what's missing? If your data is not massive, **NOTHING.**

You should use a relational database. They rock*

* - This statement has not been approved by the HBase product management committee, and neglects known deficiencies with the relational model such as poor modeling of hierarchies and graphs, overly rigid attribute structure enforcement, neglect of the time dimension, and physical optimization concerns leaking into the conceptual abstraction.



Relational DBs work well because they are close to the pure logical model.

That model is less likely to change as your business needs change. You may want to ask different questions over time, but if you got the logical model correct, you'll have the answers.



Ah, but what if you **do** have massive data? Then what's missing?



Problem: The relational model doesn't acknowledge scale.



Problem: The relational model doesn't acknowledge scale.

"It's an implementation concern; you shouldn't have to worry about it."



The trouble is, you **do** have to worry about it. So you...

- Add indexes
- Add hints
- Write really complex, messy SQL
- Memorize books by Tom Kyte & Joe Celko
- Bow down to the optimizer!
- Denormalize
- Cache
- etc ...



Generally speaking, you **poke holes** in the abstraction, and it starts leaking.



So then you hear about this thing called **NoSQL**. Can it help?



Maybe. But ultimately, it's just a different way of physically representing your same logical data model.

Some things are easier; some are much harder.

If you haven't started by understanding your logical model, you're doing it wrong.





• Scales by splitting all rows into regions



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- Each region is hosted by exactly one server



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- Scales by splitting all rows into regions
- Each region is hosted by exactly one server
- Writes are held (sorted) in memory until flush
- Reads merge rows in memory with flushed files
- Reads & writes to a single row are consistent


So what does data in HBase look like?





Table: design-time namespace, has many rows.



Table: design-time namespace, has many rows.**Row**: atomic byte array, with one row key



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Not just a bunch of bytes, dude! A k/v map!



Table: design-time namespace, has many rows. **Row**: atomic key/value container, with one row key



Table: design-time namespace, has many rows. Row: atomic key/value container, with one row key Column: a key in the k/v container inside a row



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Data Model: Brief Recap

Hold up! What about

time? ey/value container, with one row key fumn: a key in the k/v container inside a row value: a value in the k/v container inside a row





HBa

Table: Row: at Colum Timesta Value: a

But Walt...

ecap

y rows. row key a row cending ontainer



The "row" is atomic, and gets flushed to disk periodically. But it doesn't have to be flushed into just a single file!

HBase Data Model: Brief Recap



It can be broken up into different store files with different properties, and reads can look at just a subset.

HBase Data Model: Brief Recap



^f This is called "Column Families". It's ^f kind of an advanced design option, so ^r don't think too much about it yet.

HBase Data Model: Brief Recap



f f

HBase Data Model: Brief Recap





Calling these "columns" is an unfortunate use of terminology. They're not fixed; each row can have different keys, and the **names are not defined at runtime**. So you can represent another axis of data (in the **key** of the key/value pair). More on that later.



HBase Data Model: Brief Recap



They're officially called "column qualifiers". But many people just say "columns".

Or "CQ". Or "Quallie". Now you're one of the cool kids.



HBase Data Model: Brief Recap



What data types are stored in key/value pairs?

HBase Data Model: Brief Recap



What data types are stored in key/value pairs? It's all **bytes**.

HBase Data Model: Brief Recap



What data types are stored in key/value pairs? Row keys, column names, values: arbitrary bytes

HBase Data Model: Brief Recap



What data types are stored in key/value pairs?

Row keys, column names, values: arbitrary bytes Table and column family names: printable characters

HBase Data Model: Brief Recap



What data types are stored in key/value pairs?

Row keys, column names, values: arbitrary bytes Table and column family names: printable characters Timestamps: long integers

HBase Data Model: Brief Recap



One more thing that bears repeating: every "cell" (i.e. the time-versioned value of one column in one row) is stored "fully qualified" (with its full rowkey, column family, column name, etc.) on disk.



So now you know what's available. Now, how do you model things?



Let's start with the entity / attribute / relationship modeling paradigm,

and see how far we get applying it to HBase.



A note about my example: it's for clarity, not realism.

For bands & shows, there's not enough data to warrant using HBase, even if you're tracking every show by every band for all of human history. It might be GB, but not TB.



So, back to entities (boxes). With fancy rounded corners.

entity



What's that in HBase? A table, right?

table



What's that in HBase? A table, right?

Dramatic foreshadowing: not always ...





And what do entities have?

entity



Attributes!





Attributes!

For example: a "Band" entity might have a "Name" (with values like "Jonathan Coulton") and a "Genre" (with values like "Nerd Core")





Logically, attributes are **unordered**.

Their vertical position is meaningless. They're simply **characteristics** of the entity.





Attributes can be identifying.

i.e. they uniquely identify a particular instance of that entity.




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Logical models usually leave this out, and identity is implied (i.e. we just assume there's *some* set of attributes that's identifying, but we don't have to say what it is)

Attributes can be identifying.

i.e. they uniquely identify a particular instance of that entity.



Physical models refer to it explicitly, like how in a relational database model you'd label some set of attributes as being the "Primary Key" (PK).



Attributes can be identifying.

i.e. they uniquely identify a particular instance of that entity.



So in our Band example, neither Name nor Genre is uniquely identifying, but something like URL might be.



How does this map to HBase?



How does this map to HBase?

Identifying attributes (aka the "PK") become parts of the row key, and other attributes become columns.

table		
row key	key attr 1 key attr 2	
columns	column 1 column 2	



So here's a Band schema. The row key is the URL, and we have Name and Genre as attributes.





A much common pattern is to use IDs. That way, you have an immutable way to refer to this entity forever, even if they leave MySpace.





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Where do IDs come from? We'll talk about that later.



If there's just a single row key, how can there be multiple identifying attributes?

table		
row key	key attr 1 key attr 2	
columns	column 1 column 2	



Let's call that "mashing".

Which is to say, concatenation, in either a fixed-width or delimited manner, in the byte array.

band_id

url



Mashing includes several techniques:

- Fixed byte width representation
- Delimiting (for variable length)
- Other serialization (avro, etc.)



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- Fixed byte width representation
- Delimiting (for variable length)
- Other serialization (avro, etc.)

Doesn't matter how you do it; the important thing is that any byte array in HBase can represent more than one logical attribute.



If we want, we can even add types to the schema definition.

	table	
row key	key attr 1 key attr 2	byte[8] timestamp
columns	column 1 column 2	string byte[?]



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	table	
row key	key attr 1 key attr 2	byte[8] timestamp
columns	column 1 column 2	string byte[?]



HBase don't care, but we do (sometimes).



If we want, we can even add types. You could also mark things as ASC or DESC,

depending on whether you invert the bits.



	table	
row key	key attr 1 key attr 2	byte[8] timestamp
columns	column 1 column 2	string byte[?]





This is pretty textbook stuff, but here's where it gets exciting.



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(If you're astute, you'll notice we haven't talked about **relationships** yet.)



HBase has no foreign keys, or joins, or cross-table transactions.

This can make representing relationships between entities ... tricky.



Part of the beauty of the **relational** model is that you can punt on this question.

If you model the entities as fully normalized, then you can write any query you want at run time, and the DB performs **joins** on tables as optimally as it can.

(This relies heavily on indexing.)



In HBase (or any distributed DB) you don't have that luxury.

Joins get way more expensive and complicated across a distributed cluster of machines, as do the indexes that make them happen efficiently.

HBase has neither joins nor indexes.



You have two choices, if you need relationships between entities.



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• Roll your own joins



You have two choices, if you need relationships between entities.

- Roll your own joins
- Denormalize the data



Rolling your own joins is hard.

You'll probably get it wrong if you try to do it casually.

Ever implemented a multi-way merge sort by hand?



Denormalizing is fun!

But it also sucks. We'll see why.



The basic concept of denormalization is simple: two **logical** entities share one **physical** representation.



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Of course, there are lots of ways to skin a cat...



Let's look at standard relational database denormalization techniques first.



In a relational database, if the normalized physical model is this:





You could start with the Band, and give it extra columns to hold the shows:





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That's pretty obviously a bad idea, because bands can play a lot of shows.





You could also just give it an unstructured blob of text for all the shows.





You could also just give it an unstructured blob of text for all the shows.

Band_Show_Set	1	Band)
band_id show_id set order Part Of	Played By Flay	band_id name hometown genre show_list_blob	
Contains	B	ut then you	u've given up
show_id venue_id date	th (W If s	e integrity /hich might be f so, stop here.)	of your data. ^{ine.}
start_time cover_price			

You get similar problems if you try to bring all the info into the Show table.




Another solution is to focus on the junction table.





And pull in copies of the info in the other tables:





Leaving you with one table, with one row per band/show combination:

Band_Show_Set	
show_id	
set order	
band_id	
band_name	
band_hometown	
band_genre	
show_venue_id	
show_date	
show_start_time	
show_cover_price	
L	



The cons to this should be self-evident.

Updating and querying are more complicated, because you have to deal with many representations of the "same" data, which could get out of sync, etc.



But the pros are that with **huge** data, answering some questions is much faster.

(If you happen to be asking the query in the same way you structured the denormalization, that is.)



So back to HBase. How do you denormalize in an HBase schema?



HBase columns can be defined at runtime. A column doesn't have to represent a pre-defined attribute.

table		
row key	key attr 1 key attr 2	byte[8] timestamp
columns	column 1 column 2 <column n=""></column>	string byte[?] byte[?]



HBase columns can be defined at runtime.

A column doesn't have to represent a pre-defined attribute. In fact, it doesn't have to be an **attribute** at all.





A set of dynamically named columns can represent **another entity**!









If you put data into the column name,

and expect many such columns in the same row, then logically, you've created a **nested entity**.





You can scan over columns.

See: <u>hadoop-hbase.blogspot.com/2012/01/hbase-intra-row-scanning.html</u>





So we can store shows inside bands.

Which is like the denormalization we say earlier, except without the relational DB kludges.





Note!

HBase don't care.

It's just a matter of how **your app** treats the columns. If you put repeating info in column names, you're doing this.





Why is this so difficult for most relational database devs to grok?

Because

relational databases have no concept of nested entities!

You'd make it a separate table in an RDBMS, which is more flexible but much more difficult to optimize.



It's very difficult to talk about HBase schemas if you don't acknowledge this.

You don't have to represent it this way (with a nested box) but you have to at least acknowledge it. And maybe name it.



But, once you do acknowledge it,

you can do some neat things.



Nested entities can have attributes, some of which are identifying.

table		
row key	key attr 1 key attr 2	byte[8] timestamp
columns	column 1 column 2	string byte[?]
	nested	entity
	key attr 1 key attr 2	byte[1] bit[4]
	value	byte[?]



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Identifying attributes make up the **column qualifier** (just like row keys, it could be multiple attributes mashed together)





Nested entities can have attributes, some of which are identifying.

Identifying attributes make up the **column qualifier** (just like row keys, it could be multiple attributes mashed together)



Non-identifying attributes are held in the value (again, you could mash many attributes in here)

Shows is nested in Band

show_id is the column qualifier Other attributes are mashed into the value



1 table can have many nested entities, provided your app can tell them apart.

Band		
row key	band_id	int
columns	name hometown genre	string string string
	Shows Albums	
Members		



How do you tell them apart? With prefixes ...

Band		
row key	band_id	int
columns	name hometown genre	string string string
	Show	s
	Album	IS
	Membe	ers

qualifier starts with:
"s" + show_id
"a" + album_id
"m" + name



table		
row key	key attr 1	byte[8]
	nested ke	ey entity 1
	key attr 1	byte[1]
	key attr 2	bit[4]
columns	column 1	string
	nested	entity 2
	key attr 1	byte[1]
	key attr 2	bit[4]
	value	byte[?]
	nested	entity 3
	key attr 1	byte[1]
	sub nest	ed entity!
	key attr 1	byte[1]
	value	byte[?]
	nested	entity 4
	column 2	byte[1]
	timestamp	byte[1]
	value	byte[?]



• In columns

table		
row key	key attr 1	byte[8]
	nested ke	ey entity 1
	key attr 1	byte[1]
	key attr 2	bit[4]
columns	column 1	string
	nested	entity 2
	key attr 1	byte[1]
	key attr 2	bit[4]
	value	byte[?]
	nested	entity 3
	key attr 1	byte[1]
	sub nest	ed entity!
	key attr 1	byte[1]
	value	byte[?]
	nested	entity 4
	column 2	byte[1]
	timestamp	byte[1]
	value	byte[?]



- In columns
- Two levels deep in columns!

table			
row key	key attr 1	byte[8]	
	nested ke	ey entity 1	
	key attr 1	byte[1]	
	key attr 2	bit[4]	
columns	column 1	string	
	nested	entity 2	
	key attr 1	byte[1]	
	key attr 2	bit[4]	
	value	byte[?]	
	nested entity 3		
	key attr 1	byte[1]	
	sub nest	ed entity!	
	key attr 1	byte[1]	
	value	byte[?]	
	nested	entity 4	
	column 2	byte[1]	
	timestamp	byte[1]	
	value	byte[?]	



- In columns
- Two levels deep in columns!
- In the row key?!

table		
row key	key attr 1	byte[8]
	nested ke	ey entity 1
	key attr 1	byte[1]
	key attr 2	bit[4]
columns	column 1	string
	nested	entity 2
	key attr 1	byte[1]
	key attr 2	bit[4]
	value	byte[?]
	nested	entity 3
	key attr 1	byte[1]
	sub nest	ed entity!
	key attr 1	byte[1]
	value	byte[?]
	\subseteq	
	nested	entity 4
	column 2	byte[1]
	timestamp	byte[1]
	value	byte[?]



- In columns
- Two levels deep in columns!
- In the row key?!
- Using timestamps as a dimension!!

table		
row key	key attr 1	byte[8]
	nested ke	ey entity 1
	key attr 1	byte[1]
	key attr 2	bit[4]
columns	·	
columna	column 1	string
ĺ	nested	entity 2
	key attr 1	byte[1]
	key attr 2	bit[4]
l	value	byte[?]
	nested	entity 3
	key attr 1	byte[1]
	sub nest	ed entity!
	key attr 1	byte[1]
	value	byte[?]
	nested	entity 4
	column 2	byte[1]
	timestamp	byte[1]
	value	byte[?]



This is a fundamental modeling property of HBase: nesting entities.



Wait, what about Column Families?



They're just namespaces--additional vertical sections on the same entity.

table		
row key	key attr 1 key attr 2	byte[8] timestamp
family: "c	f1" column 1 column 2	string byte[?]
family: "c	f2" column 1 column 2	string byte[?]



Where column families aren't shown explicitly, let's assume there's just one.

table		
row key	key attr 1 key attr 2	byte[8] timestamp
	column 1 column 2	string byte[?]



So that brings us to a standard way to show an HBase schema:

Table: Top level entity name, fixed at design time.

Row key: Can consist of multiple parts, each with a name & type. All data in one row shares the same row key. Can be nested.

Column family: A container that allows sets of columns to have separate storage & semantics.

Column qualifiers: Fixed attributes--design-time named columns, holding a value of a certain type.

Nested entities: Run-time named column qualifiers, where the qualifier is made up of one (or more) values, each with a type, along with one or more values (by timestamp) which also have a type.

Nested versions: Because multiple timestamped copies of each value can exist, you can also treat the timestamp as a modeled dimension in its own right. Hell, you could even use a long that's not a timestamp (like, say, an ID number). *Caveats apply.*

table name		
row key	key attr 1	byte[15]
	key attr 2	timestamp
cf1	column 1	byte[15]
	column 2	timestamp
	nested entity	
	key attr 1	byte[15]
	key attr 2	timestamp
	value	byte[15]
- 10		
C12	nested versions	
	attribute 1	byte[15]
	value	bvte[15]



"But," you say, "what if I don't have all of these fancy modeling tools?"



Say it in text. XML, JSON, DML, whatever you like.

```
<key>
  <column name="band_id" type="int" />
</key>
 <columnFamily name="cf1">
  <column name="band name" type="string"/>
  <column name="hometown" type="string"/>
  <entity name="Show">
  <key>
    <column name="show_id">
  </key>
   <column name="date" type="date" />
 </entity>
 </columnFamily>
```



Text is faster and more general, but slightly harder to grok quickly.

So we'll stick with diagrams here.


Some examples are in order.



Example 1: a simple parent/child relationship Relational Schema: **Applicants & Answers**



Standard parent/child relationship. One **Applicant** has many **Answers**; every **Answer** relates to a single **Applicant** (by id). It also relates to a **Question** (not shown). SQL would have you JOIN them to materialize an applicant and their answers.



HBase Schema: Answers By Applicant



- Answer is contained implicitly in Applicant
- If you know an applicant_id, you get O(1) access
- If you know an applicant_id AND question_id, O(1) access
- Answer.created_date is implicit in the timestamp on value



HBase Schema: Answers By Applicant



- You get answer history for free
- Applicant.applied_date can be implicit in the timestamp
- More attributes on applicant directly? Just add them!
- Answers are atomic and transactional by Applicant!!



Example of rows in HBase:

ap123	name = "John Smith", q1234="foo", q3456="bar"
ap456	name = "Tom Jones", q1234="baz"
ap789	name = "A Smithee", q2345="fred", q7584="flob", q9984="fee"



Before you get too excited, remember that there are cons to denormalization.



The cons:

• Nested entities aren't independent any more.

• **NO:** "SELECT avg(value) FROM Answer WHERE question_id = 123;"

- \circ But you can still do that with map/reduce
- This decomposition only works in one direction.
 - \circ No relation chains without serious trickery.
 - Timestamps are another dimension, but that counts as trickery.
- No way to enforce the foreign key to another table.
 - \circ But would you really do that in a big RDBMS?
- On disk, it repeats the row key for every column
 - You didn't really save anything by having applicant_id be implied.
 - \circ Or did you? Compression negates that on disk ...
 - \circ ... and prefix compression (<u>HBASE-4218</u>) will totally sink this.



Example 2: dropping some database science Relational Schema: **Users And Messages**



Many-to-many relationship. One **User** sees many **Messages**, and a single **Message** can be seen by many **Users**. We want to do things like show the most recent message by subject (e.g. an inbox view).



What kind of SQL would you run on this? Say, we want to get the most recent 20 messages for 1 user.

SELECT TOP 20 M.subject, M.body FROM User_Message UM INNER JOIN Message M ON UM.message_id = M.message_id WHERE UM.user_id = <user_id> ORDER BY M.sent date DESC



Seems easy, right? Well, the database is doing some stuff behind the scenes for you:

Assuming no secondary indexes, it might:

- Drive the join from the User_Message table
- For each new record with our given user_id, do a single disk access into the Message table (i.e. a hash_join)
- Get the records for *every* message for this user
- Sort them all
- Take the top 20

No shortcuts; we can't find the top 20 by date w/o seeing ALL messages.

This gets more expensive as a user gets more messages. (But it's still pretty fast if a given user has a reasonable number of messages).



How could you do this in HBase? Try the same pattern as parent / child?

Messages By User				
user_id	byte[15]			
name email	string string			
Messag	ges_Denorm			
message subject	e_id byte[15] string			
body	string			
	essages By user_id name email Message subject body			

- 1. Because of the many-to-many, you now have N copies of the message (one per user).
 - Maybe that's OK (especially if it's immutable!). Disk is cheap.
- 2. Your statement has to do the same amount of work, but now you have to do it yourself. :(



If I **know** that I always want it ordered by date, why not store it that way?

Messages By User					
row key	user_id	byte[15]			
columns	name email	string string			
	Messages	Messages_By_Date			
	sent_date	timestamp d string			
	subj + body	y string			

- Now I can scan over messages by date until I get enough; it's O(1)
- But what if I want it by message_id again? Doh. I'm screwed, unless ...



I store it both ways!

Me	ssages By l	Jser
row key	user_id	byte[15]
columns	name email	string string
	Message	es_Denorm
	message_ subject	id byte[15] string
	body	string
	Message	s_By_Date
	sent_date message_	timestamp id string
	subj + bod	y string

Nice: updates to this are transactional (consistent) for a given user, because it's all in one row. So it's not a bear to maintain.



Which I could even do in different column families ...

M	essages By	User
row key	user_id	byte[15]
col fam 1	name email	string string
col fam 2	Messag	es_Denorm
	message subject	_id byte[15] string
	body	string
col fam 3	Message	es_By_Date
	sent_date message	e timestamp _id string
	subj + bo	dy string

(Makes sense if I don't usually access them at the same time; I only pay the cost for the query I am running.)



So, for example ...

	cf1
u123	name = "John Smith", email = "jsmith@foo.com"
	messages_denorm
	["m1234+hello my name is"="Hi john, my name is", "m234"
	messages_by_date
	"2012-05-12+m1234"="Hi john, my name is", "2012-05-13+"
	cf1
u456	name = "Tom Jones", email = "t@foo.com"
	messages_denorm
	"m1234+sales deck"="Where's that sales deck", "m234"
	messages_by_date
	"2012-05-12+m4893"="Where's that sales deck", "m4882"
	1



Or I could just use the by-date one as an "index" ...

М	essages By	User
row key	user_id	byte[15]
columns	name email	string string
	Messag	es_Denorm
	message subject	_id byte[15] string
	body	string
	Messag	es_By_Date
	sent_date message	e timestamp _id string

So I only store the subject and body once. This means I need to perform my own "join" in code.



See a theme emerging?

Relational DBs lull you into **not** thinking about physical access as much; but when you have to scale, there are hard limits.

HBase makes you think about it sooner, but gives you the tools to think about it in more a straightforward way.



Example 3: Flurry

See: http://www.flurry.com/data/

	Users			
row key device_id_hash bytes				
cf: info device_id string device info string				
cf: sessions	Session_Fragments			
	key sessio fragme fragme	on_id ent_type ent_id		
	val report	_data		

- One row per device
- Nested entity in CF "Sessions"
 - Polymorphic "fragments" of session reports
 - Map/Reduce transforms

Caveat: this is based on email exchanges, so the details may be wonky, but I think the overall point is right.



Example 4: Indexes

Example from Andrew Purtell



- Using one table as an index into another often makes sense
- Can't be transactional (without external coordination)
- So you have to keep it clean, or tolerate dirty indexes
- Note that there are several attempts to add solid general purpose indexing to HBase, but so far none have caught on.

Same caveat: this is based on email exchanges, so the details may be wonky, but I think the overall point is right.



Here are some more design patterns.



0: The row key design is the single most important decision you will make.



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This is also true for the "key" you're putting in the column family name of nested entities.



1: Design for the questions, not the answers.



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(hat tip to Ilya Katsov from the High Scalability blog for this useful way to put it; and possibly to Billy Newport or Eben Hewitt for saying it first.)



1: Design for the questions, not the answers.

(Me	ssages By	User
row key	user_id	byte[15]
columns	name email	string string
	Messag	ges_Denorm
	message subject	e_id byte[15] string
	body	string
	Messag	jes_By_Date
	sent_dat message	e timestamp e_id string



Let's be clear: this sucks big time, if you aren't 100% sure what the questions are going to be.



Let's be clear: this sucks big time, if you aren't 100% sure what the questions are going to be.

Use a relational DB for that! Or a document database like CouchDB ...



"But isn't NoSQL more flexible than a relational DB?"

For column schema? Yes! For row key structures, NO!



2: There are only two sizes of data: too big, and not too big.



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(That is, too big to scan all of something while answering an interactive request.)



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М	lessages By	User
row key	user_id	byte[15]
columns	name email	string string
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	<u> </u>	



3: Be compact. You can squeeze a lot into a little space.



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	openTSDB	
row key	metric ID	byte[3]
	base timestamp	byte[4]
	tags	
	name id	byte[3]
	value id	byte[3]
cf: " t "	data point	s
	timestamp delta	bit[12]
	flag_float	bit[1]
	reserved	bit[1]
	measurement	byte[8]



This is also important because the rowkey and CF name are repeated for every single value (memory and disk).

File compression can negate this on disk, and prefix compression will probably negate this in memory.



4: Use row atomicity as a design tool. Rows are updated atomically, which gives you a form of relational integrity in HBase!



4: Use row atomicity as a design tool.

If you made this two HBase tables, you couldn't guarantee integrity (updated to one could succeed, while updates to the other fail).

	Answer		_			
primarv kev	answer id	bvte[15]			Applicant	
columns	applicant id	byte[15]	prim	nary key	applicant_id	byte[15]
continue	question_id	byte[15]	colu	mns	name	string
	value	string			applied date	timestam
	created_date	timestamp	_			_


4: Use row atomicity as a design tool. If you make it one table with a nested entity, you can guarantee updates will be atomic, and you can do much more complex mutations of state.





5: Attributes can move into the row key Even if it's not "identifying" (part of the uniqueness of an entity), adding an attribute into the row key can make access more efficient in some cases.



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5: Attributes can move into the row key This ability to move things left or right (without changing the physical storage requirements) is part of what Lars George calls "folding".

Also, if you like this subject, go watch his videos on advanced HBase schema design, they're awesome.



6: If you nest entities, you can transactionally pre-aggregate data. You can recalculate aggregates on write, or periodically with a map/reduce job.



Practically: how do you load schema into HBase?



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Direct actuation: <u>https://github.com/larsgeorge/hbase-schema-manager</u>



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Direct actuation: <u>https://github.com/larsgeorge/hbase-schema-manager</u> Coming soon: script generation: <u>https://github.com/ivarley/scoot</u>



Thank you!

Questions? <a>@thefutureian

