

Data as software as data

scaling data science at Pinterest

Andrea Burbank



Pinterest







data is not raw material











Data as software as data

better software for better data better data for better software



Outline

- 1. No logging at all
- 2. Logs, but no insights
- 3. Incorrect data due to external factors
- 4. Incorrect data due to internal factors
- 5. Data gets messy
- 6. Correlation and causation with data
- 7. Software can make data easier





Part I: who is coming?

counting is hard









Huge growth in 2011-2012



Huge growth in 2011-2012







first name last name email address signup date & time username gender email settings



SO

first name last name email address signup date & time username gender email settings referrer landing page signup funnel pages viewed

Why such growth?

- invite-only → more desirable?
- invite-only → more homogeneous?
- viral spread through a small community?
- lots of traffic from Facebook?
- extensive press coverage?



Daily active users

From Wikipedia, the free encyclopedia

Daily active users (DAU) is one of the ways used for measuring success of an internet product, e.g., online social games.^[1] DAU measures the "stickiness" of an online product by answering the question, "How many unique users visit the site daily?" Usually the only requirement for a user to be considered "active" is that they somehow view or engage with the product. Examples of this type of usage would range from visiting the splash page of a game or commenting on a post by the game's Facebook page, to actually playing the game itself.^[2] The monthly aggregate of active users is monthly active users (MAU).









useridtspath378982308:37:03.41/pin/123/378982308:37:07.22/pin/repin/172446808:37:10.39/user/8en/877923308:37:11.97/category/art/

status	dt
200	2012-06-01
200	2012-06-01
200	2012-06-01
200	2012-06-01



userid status dt path ts 3789823 08:37:03.41 /pin/123/ 2012-06-01 200 3789823 08:37:07.22 /pin/repin/ 200 2012-06-01 1724468 08:37:10.39 /user/8en/ 200 2012-06-01 8779233 08:37:11.97 /category/art/ 200 2012-06-01

DAU: users with a request





userid	ts	path	status	dt
3789823	08:37:03.41	/pin/123/	200	2012-06-01
3789823	08:37:07.22	/pin/repin/	200	2012-06-01
1724468	08:37:10.39	/user/8en/	200	2012-06-01
8779233	08:37:11.97	/category/art/	200	2012-06-01
4980307	08:37:12.38	/email/tracking.gif	200	2012-06-01

userid	ts	path	status	dt
3789823	08:37:03.41	/pin/123/	200	2012-06-01
3789823	08:37:07.22	/pin/repin/	200	2012-06-01
1724468	08:37:10.39	/user/8en/	200	2012-06-01
8779233	08:37:11.97	/category/art/	200	2012-06-01
4980307	08:37:12.38	/email/tracking.gif	200	2012-06-01

userid path status dt ts 3789823 08:37:03.41 /pin/123/ 200 2012-06-01 3789823 08:37:07.22 /pin/repin/ 200 2012-06-01 1724468 08:37:10.39 /user/8en/ 200 2012-06-01 8779233 08:37:11.97 /category/art/ 200 2012-06-01 4980307 08:37:12.38 /email/tracking.gif 200 2012-06-01

DAU: users with certain requests

SELECT userid, count(*) FROM auth_compact_view WHERE dt="%(end_date)s" AND path not like "/email/tracking.gif%%" GROUP BY userid, dt;

DAU: users with certain requests













Is it the same users every day?



Is it the same users every day?

Or new users trying us out?



Is it the same users every day?

Or new users trying us out?

Do people come back often?
Weekly Active Users



Weekly Active Users

SELECT userid, count(*) FROM auth_compact_view WHERE dt > DATE_SUB("%(end_date)s", 7) AND dt <="%(end_date)s" AND path not like "/email/tracking.gif%%" GROUP BY userid, dt;



Monthly Active Users

SELECT userid, count(*) FROM auth_compact_view WHERE dt >= DATE_SUB("%(end_date)s", 28) AND dt <="%(end_date)s" AND path not like "/email/tracking.gif%%" GROUP BY userid, dt;



Yearly Active Users

SELECT userid, count(*) FROM auth_compact_view WHERE dt >= DATE_SUB("%(end_date)s", 365) AND dt <="%(end_date)s" AND path not like "/email/tracking.gif%%" GROUP BY userid, dt;



Yearly Active Users

SELECT userid, count(*) FROM auth_com WHERE dt >= DA AND dt <="%(end AND path not lik GROUP BY userid, dt;

iew B("%(end_date)s", 365))s" ail/tracking.gif%%"





Derived tables: DAU

INSERT OVERWRITE TABLE unique_actions
 partition(dt= "%(end_date)s", action_type=0)

SELECT userid, count(*) num_actions FROM auth_compact_view WHERE dt="%(end_date)s" AND path not like "/email/tracking.gif%%" GROUP BY userid, dt;



Derived tables: DAU

unique_actions

userid	num_actions	action_type	dt
3789823	27	0	2012-06-01
1724468	135	0	2012-06-01
8779233	1	0	2012-06-01
2987345	234	0	2012-06-01
9873001	12	0	2012-06-01
8244108	87	0	2012-06-01
4027394	43	0	2012-06-01

Who are those users?

Is it the same users every day?

Or new users trying us out?

Do people come back often?

Derived tables: WAU

INSERT OVERWRITE TABLE xd7_users
 partition(dt= "%(end_date)s", action_type=0)

```
SELECT userid,
    count(*) num_days_active,
    sum(num_actions) num_actions
    FROM unique_actions
    WHERE dt > DATE_SUB("%(end_date)s", '7)
    AND dt <="%(end_date)s"
    GROUP BY userid;
```



Derived tables: MAU

INSERT OVERWRITE TABLE xd28_users
 partition(dt= "%(end_date)s", action_type=0)

```
SELECT userid,
    count(*) num_days_active,
    sum(num_actions) num_actions
    FROM unique_actions
    WHERE dt > DATE_SUB("%(end_date)s", 28)
    AND dt <="%(end_date)s"
    GROUP BY userid;
```



How often do users visit?



How is next week different?



How is next week different?



What about next year?

```
select num weeks active, count(*) num users
from
(select x.userid, count(*) num_weeks_active
 from
 (select dt, userid from xd7_users
    where dt=2015-04-15 and event_type=0 x
 join xd7_users y
 on x.userid = y.userid
 where y.dt >= '2015-04-15'
   and y.dt <= '2016-04-16'
   and (y.dt - x.dt) / 7 * 7 = y.dt - x.dt
   and y.event_type = 0
 group by x.userid) w
group by 1;
```



What about next year?







We can count users!



Pinterest is sticky



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request logging



which requests count?



derived tables



xd7_users / xd28_users





Part I: who is coming?

counting is hard







Part II: how are we doing?

counting becomes useful



Daily Active Users





Daily Active Users



Daily Active Users















SPAM



Spam filtering










TiaKnauf7543@yahoo.com CherrieBottin23120@hotmail.com Delana64732@hotmail.com DeaneFerrill45683@hotmail.com ApoloniaChmela84292@hotmail.com GlennieBalay99738@hotmail.com DeettaGaraventa24915@hotmail.com Nita42659@hotmail.com Vernetta47911@hotmail.com Soon93012@hotmail.com BabetteEcheverri85476@hotmail.com NichelleZych24319@hotmail.com JeanetteKelton56943@hotmail.com

-I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal. -I was born botanical, with the soul of an animal.

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amy0lr amy1pv amy2gr amy2oj amy2za amy3ac amy3dd amy3hr amy3or amy4bj amy4hn

hrprgg0c0@aol.com mwxdkmgfz@aol.com llafduqq@aol.com gbxjphkp@aol.com eagufszmkml@aol.com h6ln5fnn9wm@aol.com llcrrhqedt@aol.com ljoeyanoaw@aol.com rfdklhqlj@aol.com pwhjknjlxzg@aol.com nbq4xh4s@aol.com

anna0is anna5uh anna5vc anna5yn anna6jn anna6uq anna9pb annabf2 annajz6 annamw8 annamy1

🚺 找小姐吗

Follow





same "about me" email regexes same name

rules





same "about me" email regexes same name

rules

bad images all 1 domain duplicate links

rules







File Edit Options Buffers Tools Python Help

```
class SignupGjtfhtSpammersJob(PossibleSpammerHiveJob):
    """There are many accounts with the name gjtfht fthtr."""
    _SIGNAL = 'signup_gjtfht_fthtr'
    _MULTIDATE_EXECUTE = True
    _QUERY_TEMPLATE = """
SET mapred.reduce.tasks=1;
```

INSERT OVERWRITE TABLE pinalytics.possible_spammers_data PARTITION(signal='signup_gjtfht_fthtr', dt)

```
SELECT id as userid, get_json(json, 'email') as info, to_date(created_at) as dt
FROM
db_users
WHERE get_json(json, 'first_name') = 'gjtfht'
AND get_json(json, 'last_name') = 'fthtr'
AND to_date(created_at) >= '%(start_date)s'
AND to_date(created_at) <= '%(end_date)s'
;
"""</pre>
```

class JerseyBagSignupSpammersJob(PossibleSpammerHiveJob):
 """In October 2014, thousands of accounts for fake jerseys and bags."""
 _SIGNAL = "signup_jersey_bag"
 _MULTIDATE_EXECUTE = True
 _QUERY_TEMPLATE = """
SET mapred.reduce.tasks=1;
-UU-:---F1 possible_spammers_data.py 19% L398 Git-master (Python)------















workflows	schedules	tokens				2016-05-26 06:26:55 UTC
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jobs	s list				Country	
					Search.	¥
5						
ntsJc	b	PossibleSpammersConditionCheck	CreateUniqueAppAc	tiveUsersJob	CreateUniqueAppAPIRequestsJol	CreateUniqu
			CreateOserAppAct	IVILIES JOB	ExactriveSpanOserJob	
	CreateRetentionMetric	sSASegmentedJob	CreateUniqueActionsCascadingJob	CreateUserS	StatesSAJob C	reateActivityMetricsByAppSASegmentedJob
	CreateHBaseRetentionMe	etricsSegmentInsertionJob	ExtractHiveUniqueActionSAJob	Extract	HiveUserStatesSAJob	CreateHBaseActivityMetricsByApp /
etrics	sSAJob	CopyRedShiftUniqueActionSA	ob CreateHBaseE	IventMetricsSegmentInsertionJob	CopyRedshift	tUserStatesSAJob
	final					
					(@arburbank

Signups, fall 2012



Signups, fall 2012



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data as software



data aggregation



workflows



software engineering



We can count again!











Really?



From: mobile PM To: team Subject: Have an iOS device? Check out "Featured" in the App Store ...

We launched v3.0 of the iOS Pinterest App today, which represents a *huge* amount of work from the mobile team. ... an inspired combination of iOS7's and Pinterest's visual language, the interaction model has been carefully designed ... Transitions, gestures, and information architecture were painstakingly thought through.







MacWorld: "What changed in iOS 7?"

Not that long ago, iOS apps running in the background on your iPhone or iPad were essentially stuck in suspended animation, unable to do anything (besides trigger alerts via Apple's "push notification" system) until you re-launched them and put them back on your display.

All of that changed [in October 2013] thanks to "Background App Refresh," a feature introduced with iOS 7.



http://www.macworld.com/article/3020539/apple-phone/4-ways-to-take-charge-of-ioss-background-app-refresh-feature.html







We used iOS 7's background fetch feature to get users' homefeeds ready for them before they launched the app!







"Only the /v3/feeds/home/ handler is being called in the background."

– iOS developer












userid	ts	path	status
2975021	08:36:49.32	/notifications/	200
2975021	08:37:03.41	/pin/123/	200
2975021	08:37:07.22	/pin/repin/	200
2975021	11:32:10.39	/v3/home/	200
2975021	11:32:11.97	/v3/callback/	200
2975021	11:32:13.23	/v3/user/	200
2975021	14:19:03.41	/pin/347/	200
2975021	14:19:07.22	/pin/repin/	200



not suspicious

slightly suspicious

very suspicious



not suspicious

slightly suspicious

very suspicious



"You can easily see long smears of blue, which are background hits to (mostly) v3_home_feed.



"You can easily see long smears of blue, which are background hits to (mostly) v3_home_feed. These we knew about already.



"You can easily see long smears of blue, which are background hits to (mostly) v3_home_feed. These we knew about already. But after the 3.0 app is launched, you can also see long smears of green triangles ...



"You can easily see long smears of blue, which are background hits to (mostly) v3_home_feed. These we knew about already. But after the 3.0 app is launched, you can also see long smears of green triangles ... and there are no corresponding smears of green circles before October 18.



"You can easily see long smears of blue, which are background hits to (mostly) v3_home_feed. These we knew about already. But after the 3.0 app is launched, you can also see long smears of green triangles ... and there are no corresponding smears of green circles before October 18. This makes me believe that when in the background, the app can do more than just fetch v3_home_feed.







As per our discussions yesterday, I actually confirmed iOS applications **can be launched due to a background fetch even when they are not visibly running** (not appearing in multitask tray).

... (this could explain the green streaks we were seeing).













useridtspathstatus297502108:36:49.32/notifications/200378982308:37:03.41/pin/123/200378982308:37:07.22/pin/repin/200172446808:37:10.39/v3/home/200877923308:37:11.97/category/art/200



userid path ts status app_state 08:36:49.32 /notifications/ background 2975021 200 3789823 08:37:03.41 /pin/123/ 200 active 08:37:07.22 /pin/repin/ 3789823 200 active 1724468 08:37:10.39 /v3/home/ background 200 8779233 08:37:11.97 /category/art/ 200 active

DAU

path userid ts 08:3 /notifications/ 2975021 .32 3789823 8:37 :03.41 /pin/123/ 3789823 08.37:07.22 /pin/repin/ 08:37:10.39 /v3/home/ 24468 8779233 08:37:11.97 /category/art/

status app_state background active active background active

200

200

200

200

200

Daily Active iPhone Users not DAU path userid ts status app_state 08:36.49.32 2975021 /notifications/ background 200 08.37:03.41 3789823 /pin/123/ 200 active 08:37:07.22 3789823 /pin/repin/ 200 active 08:37:10.39 /v3/home/ 1724468 200 background 08:37:11.97 /category/art/ 8779233 200 active

class CreateUniqueAPIRequestsJob(data_job.HiveJob): _QUERY_TEMPLATE = """

INSERT OVERWRITE TABLE unique_actions PARTITION(dt="%(end_date)s", action_type=100) SELECT userid, count(*) FROM auth_compact_api WHERE dt="%(end_date)s" -- exclude background requests on iOS AND (app_state not in ('background','inactive') OR app_state is NULL) GROUP BY userid, dt; **H H H**



Daily Active Users



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data as software



app behavior



changes in activity



explicit logging





Part II: how are we doing?

counting is even harder than we thought







Part III: avoiding data chaos

data, data everywhere and not a drop to drink!





app kafka raw_request_log signup raw_api_request_log event

Daily Active Users

api




















Data chaos: table proliferation



{'user id': 23987345, 'details': 'chocolate cake', 'link': 'allrecipes.com/devils-food-cake/', 'image_url': '/378340.png', 'parent_id': 109327598, 'private': false,



{'user_id': 23987345, 'details': 'chocolate cake', 'link': 'allrecipes.com/devils-food-cake/', 'image_url': '/378340.png', 'parent_id': 109327598, 'private': false, 'any_key_we_want': 'foobar',



{'user_id': 23987345, 'details': 'chocolate cake', 'link': 'allrecipes.com/devils-food-cake/', <u>'image_url'</u>: <u>'/378340.png'</u>, deprecated 'parent_id': 109327598, 'private': false, 'any_key_we_want': 'foobar',



{'user_id': 23987345,

'details': 'chocolate cake',

- 'link': 'allrecipes.com/devils-food-cake/',
- 'image_url': '/378340.png',

'parent_id': 109327598,

'private': NULL, added later; not backfilled

'any_key_we_want': 'foobar',



get_json(p.json, 'user_id') very slow



get_json(p.json, 'user_id') very slow do it once & reuse derived table userid path status dt ts 3789823 08:37:03.41 /pin/123/ 2012-06-01 200 3789823 08:37:07.22 /pin/repin/ 200 2012-06-01 1724468 08:37:10.39 /user/8en/ 200 2012-06-01 8779233 08:37:11.97 /category/art/ 200 2012-06-01



get_json(p.json, 'user_id') very slow do it once & reuse derived table userid path dt ts status 3789823 08:37:03.41 /pin/123/ 2012-06-01 200 3789823 08:37:07.22 /pin/repin/ 2012-06-01 200 1724468 08:37:10.39 /user/8en/ 200 2012-06-01 8779233 08:37:11.97 /category/art/ 2012-06-01 200 {'my_new_key': 'important_data'} **Jupdate schema;** @arburbank

Data clarity: thrift

struct PinPromotion {

- // next id: 10
- 1: optional i64 pinId,
- 2: optional i64 position,
- 3: optional binary insertionId,
- 4: optional i64 pinPromotionId,
- 5: optional i64 promoterId,
- 6: optional PinPromotionSource source,
- 7: optional i64 userId,
- 8: optional i64 advertiserPinterestId,
- 9: optional i64 gPinPromotionId,



Data clarity: thrift

- - // next id: 10
 - 1: required i64 timestamp,
 - 2: required binary actionId,
 - 3: optional ActionType actionType,
 - 4: optional i64 userId,
 - 5: optional PinPromotion pinPromotion,
 - 6: optional bool isFirstOrder,
 - 7: optional event_if.Event event,
 - 8: optional event_if.PinImpression impression,

@arburbank

9: optional string host,

Data clarity: thrift

- struct PinPromotionsActionEvent {
 - // next id: 10
 - 1: required i64 timestamp,
 - 2: required binary actionId,
 - 3: optional ActionType actionType,
 - 4: optional i64 userId,
 - 5: optional **PinPromotion pinPromotion**,
 - 6: optional bool isFirstOrder,
 - 7: optional event_if.Event event,
 - 8: optional event_if.PinImpression impression,

@arburbank

9: optional string host,

JSON vs. thrift untyped strongly typed







Data chaos

proliferation of tables

no clear data ownership

duplicated workstreams

lack of schema







data governance

- Central data repository
- Key shared tables across the org
- Deprecate unused tables
- Tests for data accuracy
- Self-documenting tables



data governance

- Central data repository
- Key shared tables across the org
- Deprecate unused tables
- Tests for data accuracy
- Self-documenting tables



software data governance

code

- Central data repository
- Key shared tables across the org
- code • Deprecate unused tables
- Tests for data accuracy
- code • Self-documenting tables





Part III: avoiding data chaos

well-structured data, well understood





what have we done so far?



added kafka logging



cloned our databases



built workflows



defined derived tables



removed unimportant requests



squashed spammers



understood our app



ignored phantom requests



created data schemas



added tests for data accuracy



And where did that get us?



Daily Active Users







Part IV: how do we grow? counting at internet scale



Remember this?



@arburbank






correlation

causation





Spectrum of certainty

Per capita cheese consumption

correlates with

Number of people who died by becoming tangled in their bedsheets



tylervigen.com

@arburbank

source: www.tylervigen.com/spurious-correlations



A/B testing



A/B testing control enabled 00000000000OOOOOOOOOO000000OOOOO00000 OOOOO00000OOOOO000000 OOOOO

A/B testing

control 00000000000000000000000 00000000000

enabled 000000000000OOOOOOOOOO $\cap \cap \cap \circ \circ$





A/B testing enabled control statistical significance

A/B testing



A/B testing enabled control novelty effects

A/B testing



A/B testing enabled control user segmentation

A/B testing



A/B testing enabled control randomization errors

A/B testing



Software tools

Doing the **right** thing should be **easy**

Doing the **wrong** thing should be **hard**

Build tools that make doing the right thing easy



Build tools that make doing the right thing easy

By showing:

- only significant differences
- novelty vs. long-term effects
- important user segments
- randomization errors
- and more ...





Part IV: how do we grow?

counting at internet scale





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data as software



leverage AB testing



scale insights with tools



Doing the right thing should be easy



Doing the wrong thing should be hard



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Counting is hard

How we got everything wrong

- No logging at all!
- Log every request as DAU
- Time-consuming request parsing
- Spam interfering with metrics
- Background fetches
 - Open browser tabs? Screensavers? Widgets?
- App identification, country attribution, ...

Data is unruly

Data governance (or lack thereof)

- ad hoc creation of kafka topics
- ad hoc creation of derived tables
- no schema enforcement (arbitrary JSON)
- ever-growing data = \$\$\$\$\$\$

Solutions:

- use thrift to enforce schema
- tools to manage & discover tables
- data retention policies



Data for scaled decisions

A/B testing

- Automate metrics collection
 - Make it easy to do the right thing
 - Show which results are statistically significant
 - Estimate population needed to measure effects
 - Show novelty and long-term effects separately
 - Make it hard to do things wrong
 - Show warnings when groups are imbalanced
 - Alert on anomalous metrics



Data for scaled decisions

A/B testing

- Culture of experimentation
 - Checklists
 - Test even when you know you'll ship
 - Have discipline:
 - State hypotheses ahead of time
 - Wait several weeks before shipping
 - Write down what you did and why, plus results
- Much more!







Thanks!

