fickr

A measurement-driven analysis of information propagation in the flickr social network

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Thomas Dackweiler

Overview

- I. About Flickr
- II. Data collection methodology and network topology
- III. Picture popularity
- IV. Evolution of pictures over time
- V. Social Cascades
- VI. Summary
- VII. Weaknesses and Problems

flickr

Founded: 2004

Acquired by Yahoo! In 2005

Photo sharing site with social networking feature

More than 4 billion photos

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I. Flickr overview – Front page

flickr[®]von YAHOO! Startseite Die Tour Registrieren Entdeo

Entdecken 💌

Sie sind nicht angemeldet Anmelden Hilfe

5

Fotostream von smichels durchsud Suchen 🔻



Brasil/Brazil 132747 fotos, 6253 Mitglieder

Mehr anzeigen...

Empfehlunden ത

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I. Flickr overview – Explore page

flickr° w VAHOO!

Home You - Organize - Contacts - Groups - Explore -

Explore



the most awesome content on Flickr We like to call it interestingness

Search

Signed in as yxcvbnm123mnbvcxy (1 new) Help Sign Out

Explore the gorgeousity by choosing a point in time

Select a month Choose -



Other places to explore include:

· a map of the world the popular tags page

- Camera Finder visiting the <u>FlickrBlog</u>, or
- most recent uploads video on Flickr

A year ago today Take a step back in time and see the most

winter

Sets A few of our favorite sets. Stories are told. interesting content from November 5, 2008 themes are developed, junk is collected ..

gennany green nonday itary jupuri torrooti macro indisto nature new newyork

night nikon nyc paris park party people portrait red sanfrancisco sky snow spain

summer sunset taiwan travel trip uk usa vacation water wedding white

Roadside Nostalgia

133 photos | From

Roadside Nostalgia,

Calcutta Photo Workshop

highwaygirl67

highwaygirl67

5 photos | From

with Cameras workshop in Calcutta for the

LUCIUS

movie: Born into Brothels"

KidsWithCameras

"Photos taken by the

BEHIND THE DUNES

92 photos | From ESOX

students during the Kids

Groups

There are loads of groups on Flickr, full to the brim with wonderful things...



"this group is intended to showcase photos of (and through) transparent stuff, delicate fabrics. screens, dirty windows and other such veils will ideally fill our photos' frames."



Tell a story in 5 frames (Visual story telling)



Night Images 68,203 members | 634,931 items

Night Images, A place to showcase images taken after the sun goes down...

Here are a few maps that we think are cool **Explore millions of** geotagged photos!

 'Rural Decay' in the U.S. South Snow in the Himalayas

Machu Pichu, Peru Cheese aficionados of Europe

Watch video tutorials on how to explore and how to geotag your own photostream or start geotagging

Explore Flickr Through Tags

or view his profile.

Where'd you take that?

Location search:

art australia beach birthday blue bw california canada canon china christmas city concert dog england europe family festival flower flowers food france friends fun company areas haliday italy ianan london man music nature new newyork

now



et your love flow - true colors More photos from Kris Kros or view his profile.

Activity Around You | In Your Groups | From your friends

- You Your Photostream | Organize | Upload | Your Account | Do More, Order Prints
- Places | Last 7 Days | This Month | Popular Tags | The Commons | Creative Commons | Search Explore
- Community Guidelines | The Help Forum | FAQ | Tools | Sitemap | Get Help Help

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I. Dissemination mechanisms

Search results:

Search engine provides content meta data (Titles, tags and description of photos)

Featuring:

Front page Explorer page

Links between content:

Links between pictures to navigate "Sets": Groups of similar pictures "Pools": Different user, but same themes

External Links:

Photos are reachable from external sites, blogs, emails,...

Social network:

Users can share contents with other users (friends, favorites,...)

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II. How are informations spread over social networks?

How widely does information propagate?

How quickly does information propagate?

What is the role of word-of-mouth exchanges between friends?

 \rightarrow Collect and analyze large-scale traces of information dissemination

Crawls of the favorite markings of 2.5 million users ... 33 million links (25% of the entire network) ... on 11 million photos ... over 104 consecutive days

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II. Data collection methodology

Collect ...

... the evolving state of the network (Part I)

... the evidence of information propagation between users (Part II)

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II. Data collection methodology

Crawl a significant subset of the network

→ Select random user and follow all of the friends links in forward direction (snowball sample)

→ Get a social network graph







II. Network topology

Flickr social (directed) graph:

- Each node = Flickr user

- Each edge = Friend link
- Outdegree = # Friends a user points to
- Indegree = Inverted direction

68% of the links are bidirectional Pearson's correlation coefficient: 0.76

II. Node degree distribution



55% nodes have just 1 outgoing link Average outdegree: 14 Highest number of outgoing links: 26,342 Maximum outdegree > maximum indegree

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II. Structural properties

Maximum path length between 2 nodes: 27

Average path length: 5.67

Clustering coefficient: (How thightly the neighbors are connected) Poorly-connected nodes: 0.05-0.10 Well-connected nodes: 0.2-0.4

II. Implications for information flow

Users with high in- and outdegree can potentially distribute informations more widely (expectation)

Most users are seperated by only a few hops \rightarrow Only short network paths

 \rightarrow Good pre-conditions for wide-spread dissemination

II. Data collection methodology

How to capture the dynamics?

Launch a complete crawl of this graph every day

 \rightarrow Newly created/removed friend links or users are recorded

II. Data collection methodology

How to collect evidence of information flow?

Get information on the favorite photos (exact timestamp) \rightarrow Recreate favorite marking events

Also known: State of the network at the time the marking took place → Filter the influencing factors

II. Problems

- This methodology does not consider deleted favorite marking
- How did the user come upon this favorite-marking?

 Network can not be manipulated: No changes possible and no tests in a controlled environment

III. Picture popularity

What are the various popularity metrics?





 \rightarrow Only a few pictures achieve high popularity and are spread widely

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III. Correlation coefficient

Examine the relationship between number of views, comments and fans:

Correlation coefficient (from -1 to 1):

- $-1 \rightarrow$ Negative linear relationship
- $0 \rightarrow No$ linear relationship
- $1 \rightarrow$ Positive linear relationship

III. Correlation coefficient

views ↔ # comments: 0.13 (not strongly correlated)

views \leftrightarrow # fans: 0.23

comments \leftrightarrow # fans: 0.60 (highly correlated)

III. Correlation coefficient

Correlation between views and fans for popular pictures:

Decreases from 0.23 to 0.21 for pictures with >100 fans and from to 0.13 for pictures with >1,000 fans

 \rightarrow Weaker correlation

Reasons:

- User find many pictures uninteresting
- Favorite-marking or comment only as registered user possible

III. Picture popularity

How widely does favority-marking propagate through the social network?

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III. Picture popularity

 \rightarrow <u>Compare</u> the most popular pictures in <u>local</u> neighborhoods with a <u>global</u> hotlist of pictures (Part 1)

→ Examine the distribution of fans as a function of their distance from the uploaders (Part 2)

III. Compare local and global hotlist (Part 1)

Pick 250 users randomly who have at least 1 photo

Distance	Min.	Med.	Avg.	Max.
1-hop	6	1,377	1,379	2,816
$\leq 2 hops$	2,785	199,330	174,100	290,671
\leq 3 hops	283,001	1,050,400	938,880	1,159,636
\leq 4 hops	880,051	1,625,482	1,563,500	1,667,054

→ Identify the top 100 pictures from their neighborhood (4hops max.) = local neighborhood III. Compare local and global hotlist (Part 1)

Compare:

Top 100 pictures based on the number of fans from that region

 \leftrightarrow

Globally popular top 100 pictures

\rightarrow Determine the "overlap"

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Region IDs sorted based on overlap in hotlists

1-hop neighborhood:

233 out of 250 local regions had no overlap between both hotlists largest overlap was 19 pictures

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Avg. overlapped pictures:

≤ 2-hops: 8
≤ 3-hops: 39
≤ 4-hops: 70

 \rightarrow Wider neighborhood boundaries \rightarrow More overlaps

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III. Observations (Part 1)

Different pieces of information are popular among the different regions

Close neighboorhod \rightarrow Pictures localized Wider neighborhood \rightarrow High overlap

 \rightarrow Information are reachable within a few hops

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III. Distance from fans to uploaders (Part 2)

- Fraction of fans that are 1, 2, 3 or more hops away from the uploader
 - Fraction of nodes that become fans of the pictures

III. Percentage of fans in k-hops distance from uploaders

# Fans	1-hop away	2-hops away	3+-hops away	
1-5	60	31	9	
6-100	55	32	13	
101-300	43	42	15	
301-500	37	46	17	
501-	32	49	19	

(3 million pictures and 10 million favorite-markings)

III. Distance from fans to uploaders (Part 2)

Problem: High differences in k-hop neighborhood sizes \rightarrow Hard to compare the distribution for different photos

Solution: Calculate the fraction of each k-hop neighborhood that became fans

How?:

Visit each user who is k-hops from the uploaders and count how many of them have marked the picture as a favorite

III. Percentage of fans for uploaders out of k-hop friends

# Fans	1-hop away	2-hops away	3-hops away
101-300	1.77	0.08	0.001
301-500	1.39	0.12	0.004
501-	1.14	0.17	0.009

Propagation of favorite marked photos is limited and photos are mostly spread in direct neighborhood

III. Picture popularity: Observations

- Different sets of photos are popular in different parts of the network
 - Fans of a picture are closely located to the uploaders
- Information do not propagate widely in this social network

IV. Evolution of pictures over time

How quickly do fans mark photos after their upload?

IV.Four different patterns of growth in popularity



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IV. Long term trends in popularity growth

Consider photos which are older than 1 year and photos which are older than 2 years

 \rightarrow Filter those pictures with more than 100 fans

 \rightarrow 5,346 and 897 photos, respectively

IV. Long term trends in popularity growth



 \rightarrow Active rise after a few days

 \rightarrow After 10-20 days: Steady linear growth

 \rightarrow Almost 40% fans acquired after the first 100 days

 \rightarrow Steady growth instead of exponential \rightarrow Different ways to get popular

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Focus on the dissemination of content via links in Flickr

Social cascade = Information exchange via word-of-mouth

Problem: No knowledge about how the user found photos \rightarrow Use of heuristics

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Definition: "User A found photo P from user B"

... if there exists a user B who is a friend of A such that:

- B also marked P as a favorite
- B included P on his favorits before A
- B was a friend of A before A made photo P his favorite

 \rightarrow Photo is propagated fom B to A via a social link



(Time X before time Y)

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2 Problems:

- Multiple friends of A could have found the picture \rightarrow A received the information from all of them
 - Uploaders can not be fan of their own photos \rightarrow They are fans by default

Use the data from the 104 days crawl

Only those photos that were uploaded during this time (entire popularity history)

 \rightarrow 10,025,185 favorite markings and 3,055,361 pictures

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How many favorites were marked by social links?

Popularity	Total	Total	Social cascades		¢
(# Fans)	pictures	fans	# Fans	Perc.	
1-5	2,704,806	4,328,609	2,197,522	51%	Г
6-100	346,870	5,121,820	2,834,704	55%	
101-300	3,502	499,870	273,596	55%	
301-500	154	54,773	27,849	51%	
501-	29	20,113	8,686	43%	
Total	3,055,361	10,025,185	5,342,357	53%	Π

→ Uploaders play an important role in the social cascades of less popular pictures

v. Peer pressure in photo marking

Favorite-marking influenced by the number of friends who already have marked that photo?

V. Peer pressure in photo marking



Number of exposure to a picture by friends

Probability of becoming a fan increases with the number of friends who already marked that picture

 \rightarrow Peers influence favorite marking

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v. Time taken for social cascade hops

How long takes it to propagate information along each hop of the social cascade?

Table: Exposure time in days prior to favorite marking# Pictures# FansMin.Med.Avg.Max.3,685190,353060140904

• 35% of fans found their favorite marked picture within a week after friends' favorite marking (not shown in table)

- 50% took over 60 days
- 140 days is the average delay

 \rightarrow It takes long time to spread across each link in the network

v. Social cascades: Observations

 \rightarrow >50% of the favorite marking through social links

 \rightarrow Individuals take a long time (3 to 5 months) to mark a photo as a favorite that was already marked by a friend

("Characterizing Social Cascades in Flickr", Meeyoung Cha, First workshop on Online social networks (2008))

Study:

2 different pictures with diverse growth pattern

Object: Look for evidence of social cascades in the growth of popularity

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Observations:

"Social cascade" group accounts for over half of new fans for both photos

→ social network plays a significant role in content dissemination

Dominance of the "social cascade" group over the "other" group switches during the two popularity surges exhibited by photo B

 \rightarrow Due to other mechanisms (i.e., linking from external sites or featuring)

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("Characterizing Social Cascades in Flickr", Meeyoung Cha, First workshop on Online social networks (2008))

Highly connected nodes are more likely to disseminate pictures

But: They are also likely to replace that pictures with a different favorite very quickly

→ High transmission rate, but short duration of "infection" Vice versa for poorly connected nodes

Most efficient: Intermediate Connectivity

VI. Overall summary

- •Most fans are only a few hops from the uploader away
- •Pictures spread slowly throughout the network
- •Even popular photos do not spread widely throughout the network
- •Over 50% of users find their favorite pictures from their friends

 \rightarrow Contradict the expectations!

Explanations? ...

VI. Explanations

Model of viral marketing:

Uploader is often the only seed who distributes a photo and beyond his neighborhood there are no further distributor

Homophily:

People who like each other's photos tend to become friends and people who are friends tend to like each other's photos

VI. Explanations

Delay:

Related to the rate at which users are exposed to the new pictures friends marked

 \rightarrow Only small number of updates of recently uploaded pictures of friends

 \rightarrow Also depends on the login-frequency

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VII. Weaknesses and Problems

Use of heuristics:

We do not know exactly which mechanisms are responsible for which users' favorite-marking

We do not know **WHEN** the user viewed a photo (We only know the total view counts per photo)

VII. Weaknesses and Problems

Rarely focus on pictures with less than 100 fans

→ Different growth pattern → Fast dissemination → Limited fan population in the early stage, dormant after the first few month

 \rightarrow No conflict with the their initial expection concerning pictures with less fans

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VII. Weaknesses and Problems

No focus on other dissemination mechanisms: "Front page", "Explore page", search results or external links

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Thank you for your attention!

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