

USC Media Impact Project Data Repository Overview

December 2016

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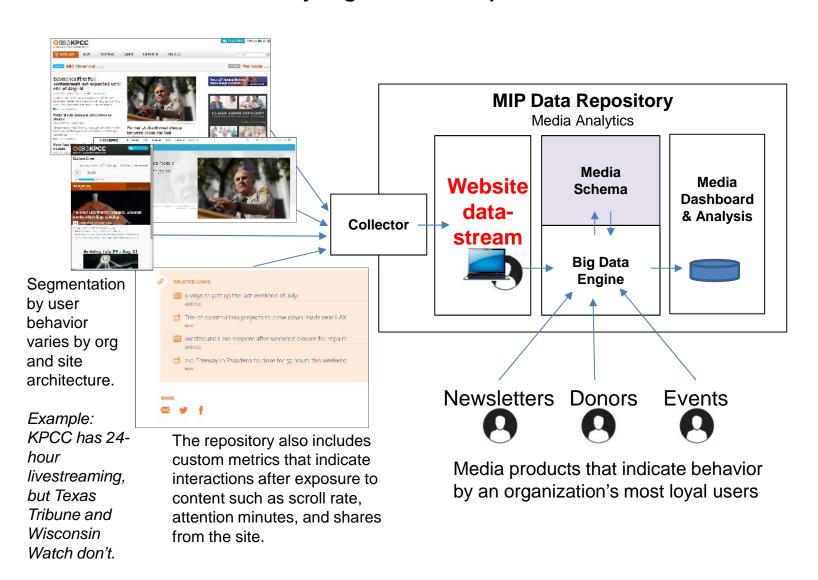
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MIP Data Repository Publicly Available Documentation

No.	Document	Description
1	Data Repository Overview (this document)	Overview of the MIP methodologies for user and story metrics for media organization managers and analysts. Description of the MIP dashboard reporting interface used by the three MIP pilot organizations to access some of the data collected in the MIP data repository.
2	Google Analytics Custom Reports	Overview of all of the custom metrics gathered with the MIP Google Tag Manager that are now available to the three pilot organizations through custom reports in the standard Google Analytics reporting interface.
3	Architecture Overview	Technical architecture overview
4	Data Repository Technical Overview	Technical architecture whitepaper
5	GitHub account	Open source code and documentation for the repository, Google Tag Manager event tracking, the dashboard reporting interface and MIP + NewsLynx.

Purpose

To understand actual people's digital media behavior across products vs. analyzing user data in product silos

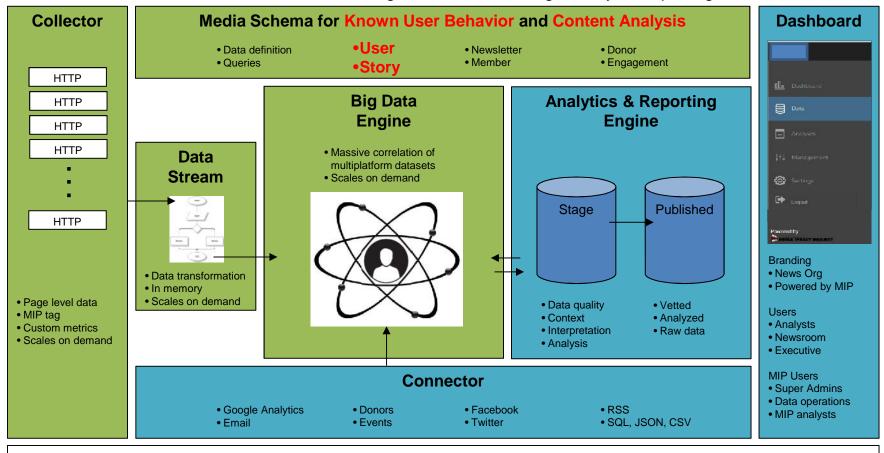


MIP Data Repository Application Architecture

News orgs have to spend more resources on grouping and classifying page URLs than other industries due to the sheer volume of content that's updated, put into context, part of a series, etc.

We have found that we can't use the raw data that come from analytics software, content management and RSS feeds, so we have to develop both automated and manual methodologies to define what a "story" is.

See page 18 for a detailed explanation of how the MIP data repository gives media organizations data about their stories that's not available through the standard Google Analytics reporting interface.



ON DEMAND MULTITENANT INFRASTRUCTURE

MIP Dashboard Reporting Interface Summary and Release Dates

No.	Section name	Section description	Page
1	Dashboard	Known user metric highlights by week	7
2	Data	 Known users: email subscribers, donors, members Stories Email newsletters MIP user vs. standard Google Analytics user counts 	8 13 21 22
3	Analyses	MIP and/or client reports	23
4	Management	System information that could affect analyses	24
5	Settings	Client-controlled user set-up	25
6	Client Management	MIP admin: Setting up clients	26
7	Data Sync Management	MIP admin: Checklist for releasing weekly data	27

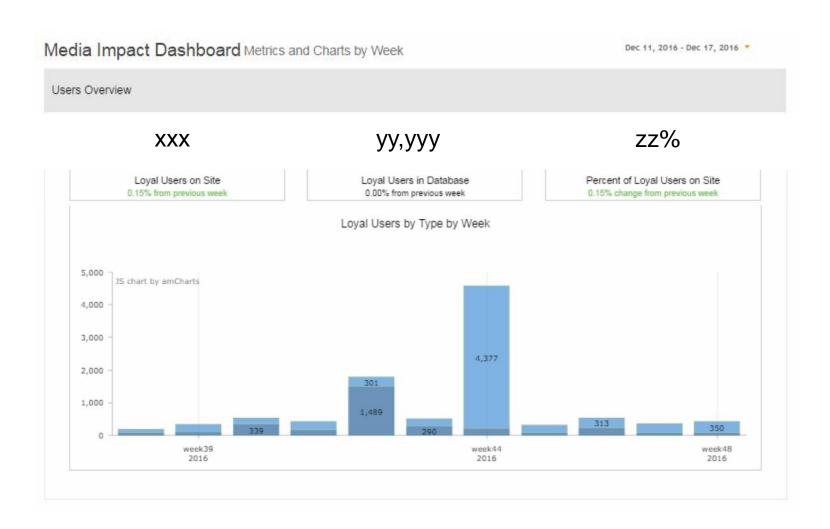
Version	Release date	Clients
0.2	July 1, 2016	KPCC (data since July 2015)
0.3	August 1, 2016	KPCC, Texas Tribune (data since July 2015)
0.4	September 1, 2016	KPCC, Texas Tribune, Wisconsin Watch (data since July 2016)
1.0	October 3, 2016	KPCC, Texas Tribune, Wisconsin Watch
	December 17, 2016	Last day of data collected

URL: http://app.mediaimpactproject.org



Section: 1-Dashboard
Tab: Users Overview

Use: How many of our most loyal users came to the site this week?



Section:	2-Data
Tab:	Users

Table: Total Known Users

Use: How many of our most loyal users came to the site this week?

Loyal users are those who have come to the site since MIP started collecting data and whose e-mail addresses are known to MIP as:

 Email newsletter subscribers only OR

2. Donors only

OR

Members who have logged in (Texas Tribune only) OR

4. Both email newsletter subscribers and donors, or other combinations of subscribers, donors and logged-in members

The **number of loyal users who come to the site each week** is a Key Performance Indicator for email newsletter performance and donor drives.

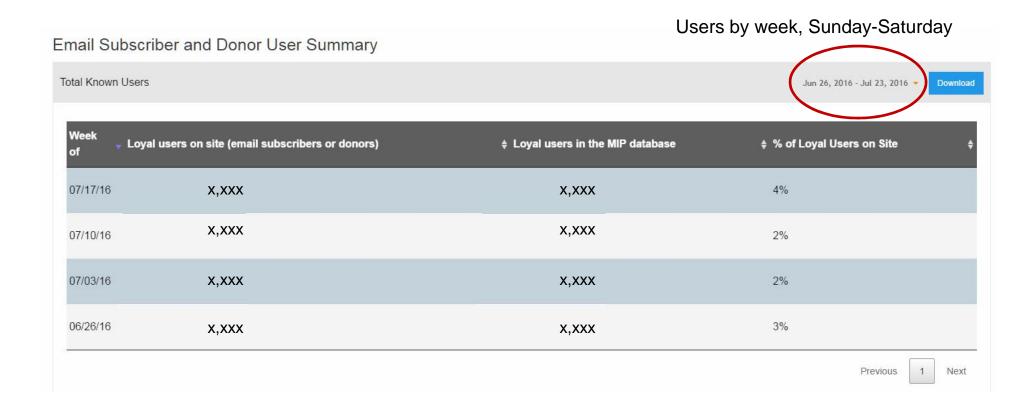
The total number of loyal users in the MIP database and percent of loyal users on site will be KPIs for total audience growth after MIP has collected data for 6-12 months.

Note: See Appendix A for more detailed explanations of how known users are identified and counted.

Section: 2-Data Tab: Users

Table: Total Known Users

Use: How many of our most loyal users came to the site this week?



Note: See Appendix A for more detailed explanations of how known users are identified and counted.

Section: 2-Data Tab: Users

Additional notes on known users

- The MIP custom tracking code allows us to identify individuals identifiable by email address. Once an individual has come to the site via an e-mail newsletter or a donation, the organization can track his/her past and future digital product behavior.
- 2. The MIP media metrics focus on identifying donors as a type of user behavior, i.e., people who donate online, with the goal of analyzing what content donors view vs. non-donors. The data repository doesn't provide metrics for analyzing donor behavior, e.g., how many times they donate, the median amounts, types of donations, etc.
- 3. Email subscriber and newsletter fields are standardized based on MailChimp (Texas Tribune and Wisconsin Watch).

Texas Tribune and Wisconsin Watch known user counts were continuously updated via direct access to MailChimp feeds. SCPR/KPCC email and donor data was from batch uploads of Oracle Eloqua data.

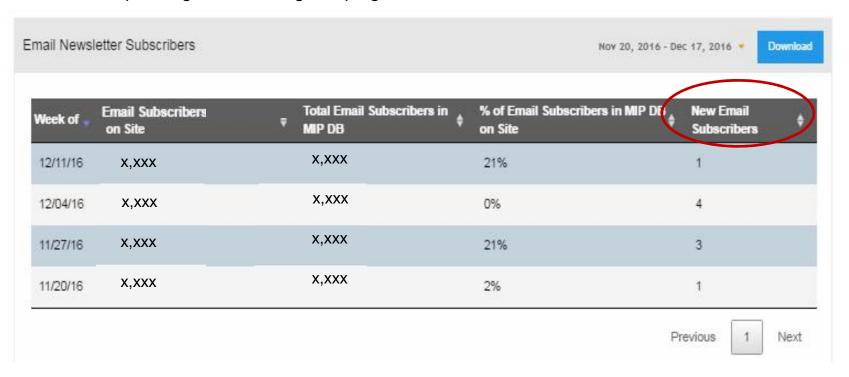
Section: 2-Data Tab: Users

Table: Email Newsletter Subscribers

Use: How many email subscribers came to the site this week?

The trend in the number of email subscribers per week is a KPI for newsletter performance. The number by specific newsletter may be available in a future Data/Newsletter tab, as will the open and click-through rates, number of clicks, and other actionable KPIs.

The number of new email subscribers per week, a possible KPI, will vary widely depending on marketing campaigns and new newsletter launches.



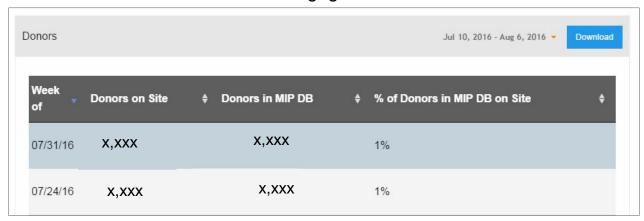
Section: 2-Data Tab: Users

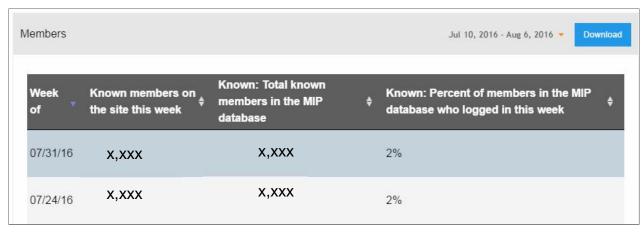
Table: Donors

Members (Texas Tribune only)

Use: How many donors came to the site this week?

The number of donors who come to the site is a KPI for analyzing hypotheses about the correlation between site/content engagement and donation revenue.





Tab: 2.2 Stories

The Stories tab has weekly numbers that include:

- Stories that are posted during the specified week, and the data for that week
- Stories that were posted in previous weeks but still got traffic during the specified week

Toggle between percentages and counts Weekly Story Performance Scroll Depth Jul 17, 2016 - Jul 23, 2016 💌 Percent Set based on each site's story architecture 25% 50% 75% 100% Started Scrolling **Article Title ♦ Total Page Views** Related Content **♦ End of Page** Scroll Scroll Scroll Chino Hills couple claims \$528 million X,XXX 95% 95% 95% 93% 90% 90% 0% Powerball jackpot Time on Article Jul 10, 2016 - Jul 16, 2016 -Percent Download 15 60 **Article Title ♦ Total Page Views** Seconds Seconds Seconds Seconds Chino Hills couple claims \$528 million X,XXX 94% 88% 74% 52% 11% 7% Powerball jackpot Differs based on the site, e.g., Texas Tribune also has "Republish" and 2016 User Interactions both "Related Content" and "Tribpedia" Total Page Share **Related Content** Click Through **♦ Tweets** ♦ **Article Title** Comments 4 Clicks Rate Rate Chino Hills couple claims \$528 million X,XXX 2 5 8 16 0% 76 1% Powerball jackpot

Tab: 2.2 Stories

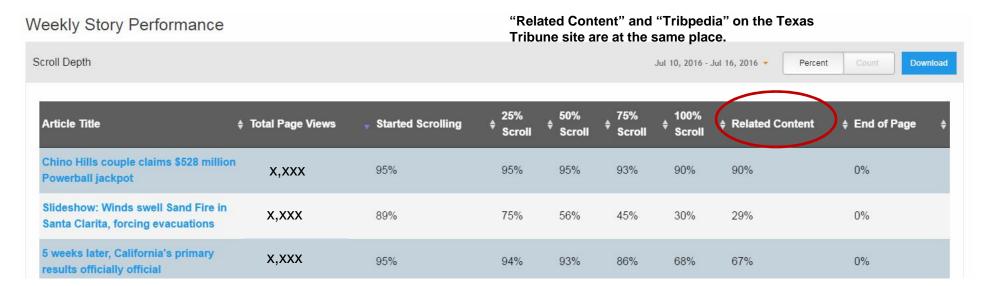
Table: Scroll depth

Use: How many pageviews involved scrolling?

Scroll depth is a proxy indicator that a user read all or part of a story. It can be used as a proxy for awareness of a particular part of a story based on the part's location.

The scroll depth for Related Content at the end of a story is combined with the Related Content clicks in the User Interactions table to calculated Related Content click rate.

MIP will be exploring hypotheses on how scroll depths, segmented by topic, can be analyzed with attention minutes, share rate and related content click rates.



Tab: 2.2 Stories

Table: Time on article, or attention minutes

Use: How many pageviews involved users seemingly paying attention?

Counting is suspended:

- If the tab isn't active in the browser
- If more than 60 seconds have elapsed since the last mouse activity on the page
- After 30 minutes on the same page (same as session timeout)

ime on Article					Jul 10, 2016 - Jul 16,	2016 Percent	Count
Article Title	≑ Total Page Views	15 Seconds	30 ♦ Seconds	45	60 ♦ Seconds	75 ♦ Seconds	90 ♦ Seconds
Chino Hills couple claims \$528 million Powerball jackpot	x,xxx	94%	88%	74%	52%	11%	7%
Slideshow: Winds swell Sand Fire in Santa Clarita, forcing evacuations	x,xxx	85%	71%	60%	50%	19%	13%
5 weeks later, California's primary results officially official	x,xxx	93%	77%	58%	42%	12%	7%

Google Analytics standard time on site (or session duration) is based on calculating when a page is sent from the server. It doesn't capture time spent on the last page a user got, or the time spent during sessions with only one page view (i.e., a bounce).

Tab: 2.2 Stories

Table: User interactions – comments, shares, related content clicks

Use: How much interaction did a story have?

We assume that users who click on "email," "Twitter," "Facebook" and other action buttons at the top and bottom on a story are more likely to have read the story than those who shared it from outside the site.

User interactions will vary greatly by topic and whether there are related stories, e.g., a user may comment on the second story he/she reads. Thus, we recommend that user interactions be analyzed by topic, story series, story format and other variables. Site-wide medians and rankings shouldn't be KPIs.

Also, each interaction type should probably be weighted differently. As they need to input email addresses and messages, users who email a story could be more highly engaged than users who share a story via Twitter or Facebook.

ser Interactions							the site, e. t" and "Tri	g., Texas Tribune I	- Jul 16, 2016 Downle
Article Title	Total Page	▼ Comments	Email Shares	tweets	FB Shares	Total Shares	Share Rate	Related Content Clicks	Click Through
Chino Hills couple claims \$528 million Powerball jackpot	x,xxx	2	5	1	8	16	0%	76 Will be zero o	1% or null if
Slideshow: Winds swell Sand Fire in Santa Clarita, forcing evacuations	x,xxx	0	45	9	16	0	0%	the article did ¹ the Related C table on the p	ontent

The scroll depth for Related Content at the end of a story is combined with the Related Content clicks in the User Interactions table to calculated Related Content click rate.

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Tab: 2.2 Stories

The Stories tab has weekly numbers that include:

- Stories that are posted during the specified week, and the data for that week
- Stories that were posted in previous weeks but still got traffic during the specified week

Not in the dashboard reporting interface, but available via custom queries or reports:

- Total traffic/metrics per story over a specified time period. This report would include:
 - Stories that were posted during the specified time period
 - The cumulative traffic/metrics per story from the time the story was posted to the date the report is run.

Note: This type of cumulative report is similar to what's available by post in Facebook Insights.

Example: The Texas Tribune needs to report on the impact of its second quarter 2016 higher education coverage. The report would include:

- Stories that were posted April 1, 2016 through June 30, 2016
- Pageviews and other metrics per story from April 1, 2016 through July 23, 2016, when the report was run.

Trending by week for a story or group of stories

This report would be used to track the impact of a story in response to external and internal events, e.g., explainers, stories in a series.

Tab: 2.2 Stories

Definition of a "story"

- The number of stories is essential to analyze content strategy and is often required for funder reporting.
- However, each organization's site architecture and analytics tagging strategy requires its own story counting methodology for URLs vs. page titles.
 - Comments pages in the Texas Tribune are separate from the story and have different URLs.
 On KPCC, comments are on the same page/URL as the story.
 - The MIP Google Tag Manager code is not included on all pages of a site. For example, it is not included on Texas Tribune app pages, so those stories and data aren't in the dashboard.
- In both the MIP data repository and in Google Analytics, pageview reports are based on unique URLs. Selecting page title or adding page title as a secondary dimension results in getting different data. The number of rows in a pageview report is usually used as a count of the total number of articles, as each unique URL usually only has one page title.
- A media site's URL and page title structure often includes too many irregularities for
 organizations for the number of rows to be an accurate count of the number of articles, or for
 any one row to include an accurate aggregation of the metrics for any one story. The
 irregularities are usually due to media organizations updating a page multiple times a day, with each
 update sometimes generating different unique URLs and/or page titles for the same article.
 - o In the standard Google Analytics reporting interface, Google Analytics will aggregate pageviews and other metrics for a site that exceeds 50,000 URLs a day. Stories will be listed as "(other)" rather than by their URLs and page titles. This is a known issue with the Texas Tribune.
 - MIP data repository reports show all URLs and page titles, regardless of whether a site exceeds 50,000 URLs a day.

Section: 2-Data Tab: 2.2 Stories

Definition of a "story" (continued)

- A story that's updated or is in a different language may have a different page title but the same URL. For example, all of the following page titles have the same URL:
 - Lawsuit Won, Texas Moves to Cut Therapy Programs
 - o Lawsuit Won, Texas Moves to Cut Children's Therapy Programs
 - o Demanda Won, Texas mueve a recortar los programas de terapia
- Similarly, the Interactions report can include data in multiple rows that need to be aggregated. For example, the number of Facebook Recommends for an article on the SCPR/KPCC site may be in three different rows:
 - Facebook : recommend
 - o N/A www.scpr.org : N/A Facebook
 - o N/A www.scpr.org : N/A Facebook ã§ã, ·ã,§ã,¢
- The Stories tab lists articles by page title, and each page title clicks to a URL. Thus, the tab
 has rows with duplicate stories, and can't be used to get the count of stories or to rank the
 stories. However, the downloaded report includes both page titles and URLs, with each row
 being a unique combination of page title and URL.
 - To get an accurate story counts and metrics by stories, organizations can download the story report, group together the stories, and aggregate the data.
 - In a test of three months of Texas Tribune story data, we found that:
 - A report of about 3,000 rows actually represented only about 1,000 stories.
 - After trying out one algorithm, there were no obvious consistent patterns to the irregularities, so the groupings of rows may need to be done manually. See Appendix B for a detailed breakdown of the irregularities.

Section: 2-Data Tab: 2.2 Stories

Implications for media metrics

- Defining a "story" is essential for all media metrics operations and research, as the raw storylevel data is incomplete and thus leads to underestimating a story's impact and mispresenting its relative importance to other stories.
 - Coding by topic, part of series, story format and other metadata should not be done until the row data is aggregated.
 - Medians by week and other descriptive statistics based on the raw data are probably irrelevant.
 - Combining methodologies will get more complicated if we add datasets from other analytics systems (e.g., MailChimp), content management systems and RSS feeds.
- The root problem is not in Google Analytics and other digital analytics tool but rather in each organization's site architecture and in its investment in the resources needed to maintain a systematic, consistent URL and page title assigning process across its content management system and all of the digital platforms it controls site, email, mobile apps, etc.
 - Media organizations can get far more from the standard Google Analytics reports with standardized, canonical URLs and page titles. For example, summarizing content performance by category could be a simple query rather than a process that involves applying a taxonomy manually after data is collected.
 - Developing best practices site architecture maintenance processes has operational benefits beyond media metrics, such as optimizing a site for search engines (SEO).

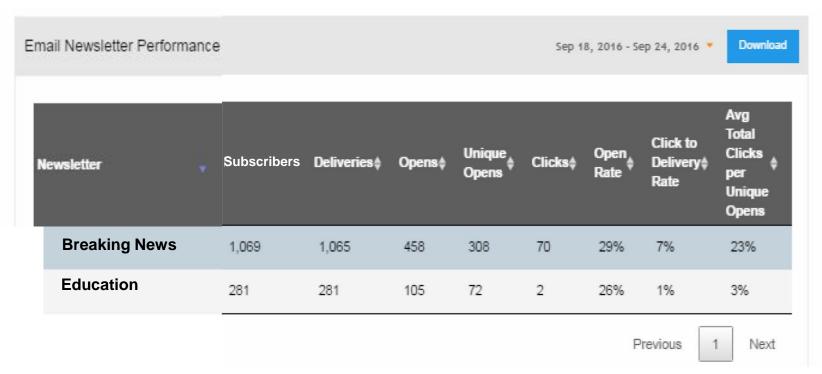
Tab: 2.3 Newsletters

Table: Email Newsletter Performance

Use: How much traffic did email newsletters send to our site? From whom?

The click per unique open rate by type of email newsletter is a KPI that can lead to more in-depth analyses of:

- What stories appealed to an organization's most loyal and known users, email newsletter subscribers
- The number and type of known users who click on an email newsletter could also be a KPI.



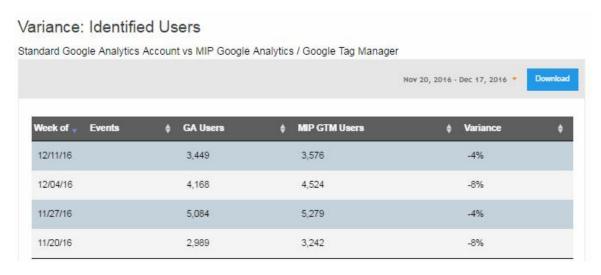
Tab: 2.4 Data Quality

Table: Variance: Identified Users

Use: How does the number of MIP users differ from the number in an organization's standard Google Analytics account?

The MIP Google Tag Manager code is on most but not all of an organization's pages. This table shows the variance in identified users from week to week. A variance higher than 10 percent may indicate a problem with the MIP GTM code.

Or, it may be due to high usage of a section of a site where the MIP GTM code doesn't exist. For example, the MIP GTM code is not on the Texas Tribune app pages, or sometimes there was a lag in putting the MIP GTM code on a new section of the site.



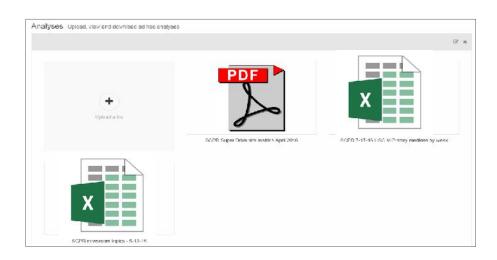
Section: 3-Analyses

Use: What analyses are available? What taxonomies should we use?

The primary purpose of the Data section is to select and download data for analysis outside of the dashboard.

The purpose of the Analyses section is for MIP and the organization to post:

- analyses that answer specific questions using data from multiple sources
- files such as taxonomies and weekly medians



Custom Texas Tribune higher education report compiled with MIP custom metrics and topic coding

		Reach			"Read	" and re	and responded		Shares after exposure			re to stori	e to stories on the site	
	(- 0		Percent	Clicks	Percent of							
			Percent of	Pageviews	of PVs	to	clicks to				Shared	Total		
	No. of		Total	scrolled	scrolled	related	related	No. of			on	shares	Percent of	Clicked on
Topic	🕌 stories	Pageviews	Pageviews	("read")	("read")	stories	stories	comments	Emailed	Tweeted	Facebook	from site	shares	"Republish"

Section: 4-Management

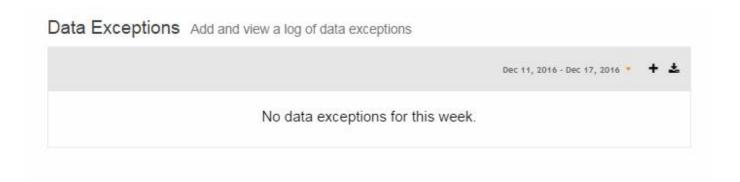
Table: Data exceptions

Use: Data quality caveats for analysts

This section is a manually updated log of any MIP or client issues that would affect the analysis and interpretation of the MIP data repository data.

Examples:

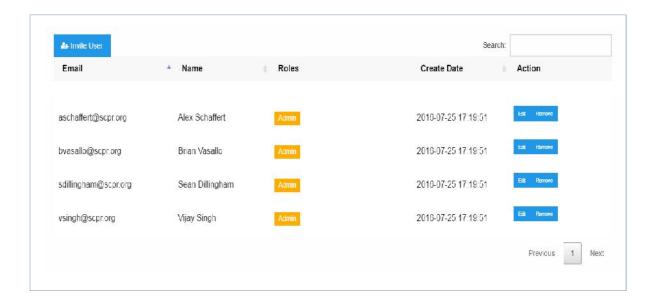
- January 19-28, 2016: GTM tracking code was removed from all but one page. This resulted in nearly no data being collected from most of the website.
- March 23, 2016: No comments were tracked due to a JavaScript race condition where the comment tracking code would not fire properly. Fix was added to the GTM update on March 24, 2016.

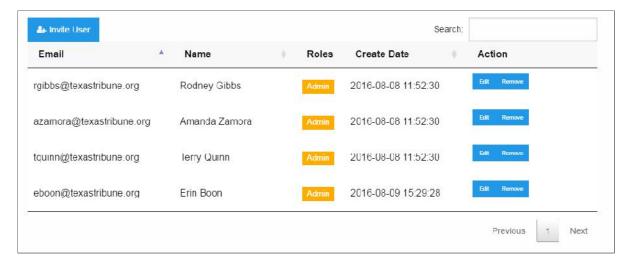


Section: 5-Settings

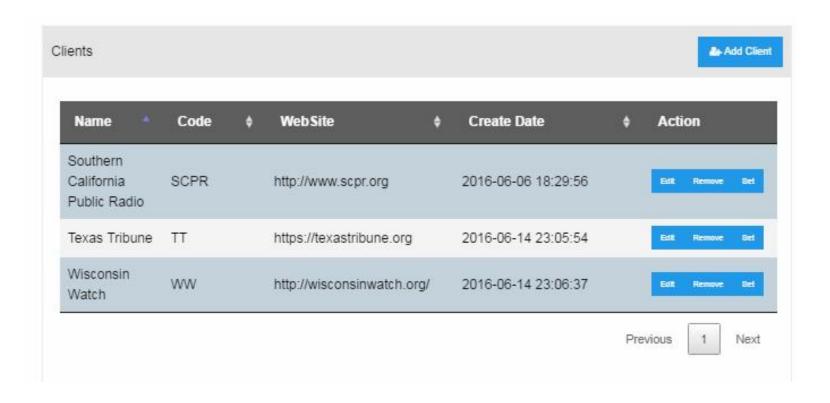
Use: Client-controlled user set-up and editing

- An organization can sign up an unlimited number of users.
- Hundreds of users across the three MIP pilot organizations can be on the system simultaneously.
- Each organization sees only its own users in this Settings section.



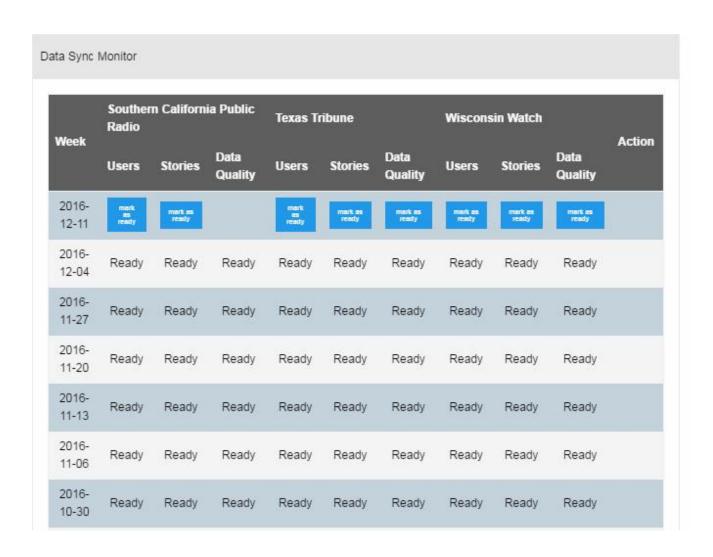


Section: 6-Client Management (MIP admin)



Section: 7-Data Sync Management (MIP admin)

Each data section was checked by an analyst each week before being released to the client.



Appendix

- A. Known User Methodology
- **B.** Texas Tribune URL and Page Title Analysis

Appendix A-1: Identified vs. Known Users – SCPR/KPCC Example

	of unique Eloqua customer code	es (ELQ)				
Click number	CID	ELQ	Date of click	Time of click	E-mail address from Eloqua database	Name from Eloqua database
1	C6748	E2738	Monday	9:00 a.m.	Jean@gmail.com	Jean Smith
2	C9854	E2738	Tuesday	9:00 a.m.	Jean@gmail.com	Jean Smith
3	C0987	E2738	Wednesday	3:00 p.m.	Jean@gmail.com	Jean Smith
4	C0987	E5934	Thursday Tuesday	4:00 p.m.	Jean@scpr.org	Jean Smith Joan Jones
5	C2637	E6473		9:05 a.m.	Joan@yahoo.com	
6	C2637	E6473	Tuesday	10:05 a.m.	Joan@yahoo.com	Joan Jones
	Number of Identified Users			4		
	Number of Known Users			3		
An e-mail newsle	tter subscriber who clicks from a n	ewsletter from	multiple devices will ha	ave multiple unique	Google cookie IDs	but only one
Eloqua ELQ= cod	le.					
Eloqua ELQ= cod	le. C6748 (Home PC, Home e-mail)	E2738	Monday	9:00 a.m.	Jean@gmail.com	Jean Smith
Eloqua ELQ= cod 1		E2738 E2738	Monday Tuesday	9:00 a.m. 9:00 a.m.	Jean@gmail.com Jean@gmail.com	Jean Smith Jean Smith
Eloqua ELQ= cod 1 2	C6748 (Home PC, Home e-mail)				- 1 Table	
Eloqua ELQ= cod 1 2 3	C6748 (Home PC, Home e-mail) C9854 (Phone, Home e-mail)	E2738 E2738	Tuesday Wednesday	9:00 a.m. 3:00 p.m.	Jean@gmail.com Jean@gmail.com	Jean Smith
Eloqua ELQ= cod 1 2 3 One person with	C6748 (Home PC, Home e-mail) C9854 (Phone, Home e-mail) C0987 (Work PC, Home e-mail)	E2738 E2738	Tuesday Wednesday	9:00 a.m. 3:00 p.m.	Jean@gmail.com Jean@gmail.com	Jean Smith
Eloqua ELQ= cod 1 2 3 One person with	C6748 (Home PC, Home e-mail) C9854 (Phone, Home e-mail) C0987 (Work PC, Home e-mail) two different e-mail addresses in the	E2738 E2738 e Eloqua datal	Tuesday Wednesday pase will be two known	9:00 a.m. 3:00 p.m. n e-mail newsletter	Jean@gmail.com Jean@gmail.com subscribers.	Jean Smith Jean Smith

Appendix A: Loyal User Calculations – SCPR/KPCC Example

Reference	F How many e-m	G nail subscribers	H came to the sit	THIS WEEK?	J low many were	K e (1 new subscrib	ers and/or (2 nev	to MIP?	N	0	do we	Q e-mail subscribers have in the MIP atabase?	R Calculated fie.d	S E-mail newsletter clicks per week
ė G	Identified: subscribers already in MIP database who came to the site this week	Known: Subscribers already in MIP database who came to the site this week	Identified: Subscribers who came to the site through an e-mai this week for the first time since MIP started collecting data	Known: Subscribers who came to the site through an e-mai this week for the first time since MIP started collecting data	Identified: Nev subscribers this week who also clicked on an e- mail this week		Identified e- mail newsletter subscribers THIS WEEK	Known e-mall newsletter subscribers THIS WEEK (unique ELQs)	identified: New e mail subscribers this week	e- Known: New e- mall subscribers this vreek	e-mail newsletter	Known: Total fied number of known e-mail newsletter s in subscribers in the MIP database	Known: Percent of subscribers in the MIP	The number of pageviews with an ELCtrackID-code from un quo Coogle cookie IDs (weekly urique users) that also have an Eoqua ELQ=code
	Should be less than P	Should be less than Q	May vary widely	May vary wde y	Should be less than N	s Should be less than O	F+H+J=L Should be compared to the number of	Should be less than L and equal G+I+K=M. Should be less than or equal to the total number of subscribers in the Eloqua database.	be equal to	ubscribers should o Known new cribers?	P should equal previous week plus H + N	Q should equal P previous week G plus I + O	0.000	Should be compared to the number of sessions coming from e-mail in Google Analytics Acquisition/Chainels/Fmail/I an ding Page-Campaign.
										1				
eference	-	U		V	W	X	Y	Z		AA	ВВ	cc	DD	EE
efe <mark>r</mark> ence	-	U V donors came			w	х	Y	How man	ny donors do ne MIP databa	we have se?	BB Calculated	How many known	DD How many kn individuals a the MIP datal	own re in
deference	-	onors Known: Dale already in database	onors Identif MIP who do who the site first tin MIP sta	ied: Users Knomated on who ifor the the since first inted MIP	vn: Users donated on ite for the	ldentified	Known donors on the site THIS WEEK	How man	d: Known the in the in the	we have se? fin: Total is donors in MIP is in the second	Calculated	How many known individuals came to the site THIS WEEK? Known: Total known donors and/or e-mail newsletter subscribers	How many kn individuals a	own re in base? Calculate field I Known: Percent of known individual in the MIP

Each unique e-mail address in the MIP database should match to an e-mail in the Eloqua e-mail and donor databases. Unique identifier is e-mail ad dress, so manual clean-up/merge-purge would be needed to reduce the number of duplicates/triplicates.

Texas T	ribune sto	ries in the MIP	data repository						
Stories	posted Api	ril 1-June 30, 2	016						
	Does no	t include stories	posted prior to April 1, 2016	that got pa	geviews April 1-June 30, 2016				
Goal	How ma	ny stories wer	re posted?						
	13210000		ory has one unique URL and o	ne corres	conding page title.				
			be able to see a clean page title and then click on a link to go to the story.						
	A.	Thus, we ran	a report with both page title	e and URL	Each row is a unique combination of page title and URL.				
		Report query:	What URLs first appeared du	ring secor	d quarter?				
	B.	However, eac	h unique URL has one or m	any page	titles, for various valid operational reasons.				
	Row ID	Page Title ID	Page Title	URL ID	URL				
			Abbott to Release Book on	137	CONTRACTOR OF THE STATE OF THE				
	52	22	Life, Convention of States	13	/2016/04/04/abbott-writes-book-life-constitutional-convention				
	53	23	Abbott Writes Book on Life, Constitutional Convention	13	/2016/04/04/abbott-writes-book-life-constitutional-convention				
	54	24	Abbott Writes Book on Life, Convention of States	13	13 /2016/04/04/abbott-writes-book-life-constitutional-convention				
		Thus, to get an	accurate story count, you sh	nould be ab	ole to run a report based on unique URLs.				
			eport useful, the TT will need to						
		1	Select which of the duplicate	page titles	should be marked as the primary page title.				
		1	This can be semi-automated	but needs	to be reviewed and finalized manually.				
		2	Combine the pageview and o	other data.					
			This can be automated.						
		Some of the duplicate page titles may only have one or two pageviews, but we don't know if these are the rows the							
			내가 하지 않는 내가 있는 것이 없었다. 그렇게 하셨다면 하는데 그리고 있는데 그렇게 하는데 되었다.	determine f so, repor	if rows with less than x pageviews have zero comments, shares is could exclude these rows, and there would be fewer duplicate				

C.	There are "du	plicate" URLs						
	1	This may be a problem with t	he way Go	ogle Analytics reports URLs.				
	2	This perhaps can be fixed wi						
	3	This may be a problem with t						
Row_ID	Page_Title_ID	Page_Title	URL_ID	URL				
2758	8	Page Not Found	789	/2016/06/10/citing-ong				
2759	8	Page Not Found	789	/2016/06/10/citing-ongoing-issues-				
2760	8	Page Not Found	789	/2016/06/10/citing-ongoing-issues-state				
2761	8	Page Not Found	789	/2016/06/10/citing-ongoing-issues-state-				
2762	8	Page Not Found	789	/2016/06/10/citing-ongoing-issues-state-scraps-=				
2763	8	Page Not Found	789	/2016/06/10/citing-ongoing-issues-state-scraps-staar				
2764	1460	Falta sÃ-mbolo del Estado Ups para el desguace de la escuela Puntuaciones de prueba	789	/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				
2765		Falta Ups sÃ-mbolo del Estado para el desguace de la escuela Puntuaciones de prueba		/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				
2766	1462	Foul Ups Prompt State to Scrap School Test Scores	789	/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				
2767	3	Login to Your Account	789	/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				
2768	8	Page Not Found	20100	/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				
2769	155	Null		/2016/06/10/citing-ongoing-issues-state-scraps-staar-scores				

D.	Some "duplica	ate" page titles have nonwo	ords.			
	1	We can automate removing	nonwords t	that are repeatedly and consistently used.		
Row_ID	Page_Title_ID	Page_Title	URL_ID	URL		
59	25	Analysis: Texas Government's Shrinking Financial Buffer	14	/2016/04/04/analysis-texas-govts-shrinking-financial-buffer		
60	26	L1 - Analysis: Texas Government's Shrinking Financial Buffer	14	/2016/04/04/analysis-texas-govts-shrinking-financial-buffer		
228	101	54.7 MB: UT Steps Up Security After Body of Student Found on Campus	63	/2016/04/07/ut-steps-security-after-body-student-found-campu		
229		58.3 MB: UT Steps Up Security After Body of Student Found on Campus		/2016/04/07/ut-steps-security-after-body-student-found-campu		
230		61.7 MB: UT Steps Up Security After Body of Student Found on Campus		/2016/04/07/ut-steps-security-after-body-student-found-campu		
231		62.0 MB: UT Steps Up Security After Body of Student Found on Campus	90000	/2016/04/07/ut-steps-security-after-body-student-found-campus		
232	105	72.7 MB: UT Steps Up Security After Body of Student Found on Campus	63	/2016/04/07/ut-steps-security-after-body-student-found-campu		

E.		하게 하는 그를 잃다. 그들은 이 사람이 얼마가 하면 하면 나가 되는 그래? 나를 하게 하다 하다 하다.		ng completely automated matching. Only some of the etitles would need to fixed manually.
Row_ID	Page_Title_ID	Page_Title	URL_ID	URL
92		Emails Show Abbott\u2019s Involvement at Child Welfare Agency	24	/2016/04/05/emails-show-abbotts-involvement-child-welfare- agen
93	38	Emails Show Abbott's Involvement at Child Welfare Agency	24	/2016/04/05/emails-show-abbotts-involvement-child-welfare- agen
F.	Some URLs have no page titles, or are null.			
	See the example in C. Duplicate URLs.			
	Need to analyze past data to determine if rows with null page titles have less than x pageviews and less than x number of zero comments, shares and other custom metrics. If so, reports could exclude these rows, and there would be fewer duplicate page titles to manually mark and combine.			
G.	Some URLs and recurring page titles don't give any information on the story topic.			
	1	This hampers tagging stories and adding other metadata.		
	2	This could be corrected with site design and would be better SEO.		
Row_ID	Page_Title_ID	Page_Title	URL_ID	URL
84	35	The Bookshelf: April 6, 2016	21	/2016/04/05/bookshelf-april-6-2016