

# Wob

# Anonymous geo-based messaging

David Hao, Kyle Meade, Kat Thambipillai, Jonathan Stroz, Julien Lin, Sophia Yang, Ehsan Merati, Graham Carkner

### **Problem Statement**

There is currently no social media platform that focuses on connecting members of a geographic community.

Secondarily, people value their data usage and privacy online, now more than ever.





Wob is the solution.





# Wob is the solution.

BTW, this is Wobbly.





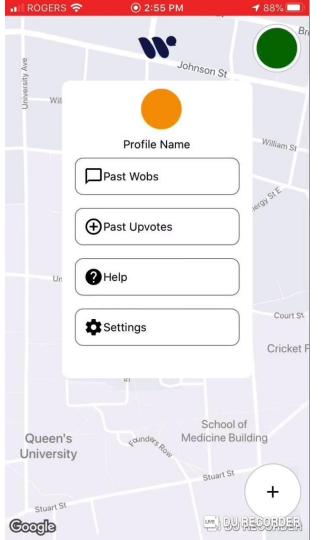
# What is Wob?

An anonymous geo-based messaging platform with an emphasis on user safety and community building.





# Let's see a demo!





### The Wob Team



# Wob fills an unoccupied gap in the market.



Allows users (and their data) to stay anonymous and safe



Allows users to be candid, sparking more interesting discussion



Everyone has equal say - no follower count





### **Competitor Analysis**

# Wob's unique features creates differentiation

Platform Feature

**Benefit Realized** 

Anonymity

More likely for users to share their opinions

Engagement

More relevant, candid, interesting discussion

Geo-Based

Rich, high-context content for users

Relevancy

More relevant content increases retention



### **Competitor Analysis**

# Safety is a key feature that sets Wob apart

Anti-bullying/harassment often lacking on other anonymous social media sites

Our methods to ensure safety

ML Filter

First defense filter applied to all WOBS

Community Filter

Users take an active role in the WOB community

**Partnerships** 

Partnerships with local authorities (e.g. universities, police)



### **Market Analysis**

# Students comprise our core target market



**Students** 

Highest level of potential engagement, campuses make for perfect social space

1.8M

Target Market

Canadian University Students.

Ample market size to generate significant network effect



### **Market Analysis**

# Young professionals are a secondary target



Young professionals

Gossipy setting, social ladder defined by information holders, similar collegiate environment

17.2M

Target Market

Young Professionals in the US

A fruitful market to capture once foothold on students is secured



So how do we make money?





## Advertisements fuel our business model

Target Strategy: Free-to-use platform powered by Advertisements

Location based advertisements that **do not obstruct content flow.** Appear as Wobs

Advertising platform that focuses on **local businesses and community**, allowing them to **connect with local consumers**.



# Pricing plan is customizable to client needs

### Base

#### Features:

- 3 day run time
- No prioritization

Cost per Click

\$0.25

### **Run Time**

#### Features:

 Ability to run ad for more than 3 days

**Additional Cost per Click** 

\$0.01 per Day

### **Prioritization**

### **Features:**

 Ability to control location and visibility of ad

Cost per Click

\$0.32



# Targets for Canada are achievable in timeline

\$800k

60%

350K

Revenue by 2025

Of Ontario Students

Users by 2025



# Here's how we're going to get to our targets





# Sponsored messages come from relevant clients

### **Local Clubs & Attractions**





### **Local Restaurants**





Advertising platform that focuses on **local businesses and community**, allowing them to **connect with local consumers**.



# Stone City Ales validates our model

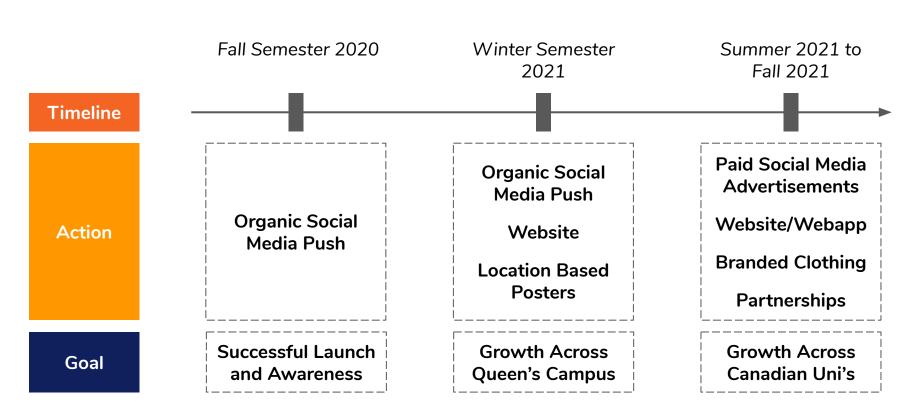


"The biggest challenge with marketing to university students is that they don't know where we are, and what we offer. It's hard for them to learn about our special releases on tap."

"There isn't an engaging way of bringing them in."



# Sample customer acq. strategy for a campus



### For Queen's University

# **Customer Acquisition Strategy**

### **Digital**

#### **Mediums:**

- Queen's Subreddit
- Facebook Groups
- Snapchat Geofilters

### Example:



Sitting in Stauffer trying to procrastinate with the hundreds of other students around you? Resting in between sets at the ARC wondering what the gym drama is? Sitting in the lecture hall wondering if anybody else gets wtf is going on? Post and hear anonymously on Word on the Block

### **Experiential**

### Types:

- Laptop stickers
- Phone grips
- Normal stickers

### Example:



### Contextual

### **Mediums:**

- Location-Based Posters
- Queensu confessions

### Example:

Hit a PR? Want to shout out a friend making mad gains? Want to elongate your rest?

Hear the Word on the Block

# State of development and next steps





### **Product Analysis**

# Core MVP features fulfill base functionality

### **Post Wobs**

Post "Wob" comments at your location whenever you want. Get anything off your chest, or don't fear to be judged

### Wobs Near Me

See what people in your community are thinking about around you

### **Voting & Comments**

Feel rewarded from credit given by others, downvote away bullies, and share your appreciation for a post.



### **Product Analysis**

# Future product roadmap expands on value prop

### Ad Integration

Let businesses purchase ads that can show up on the map.

### Ai-based filtering

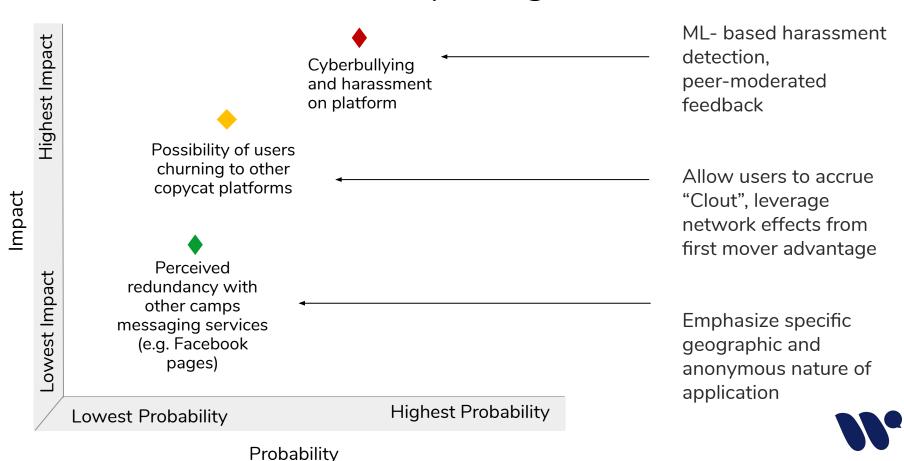
Leverage machine learning to filter out troll comments before they have the chance to be seen by the public

### Passport feature

Have the option to view posts at other places around the world, beyond just near the user's location



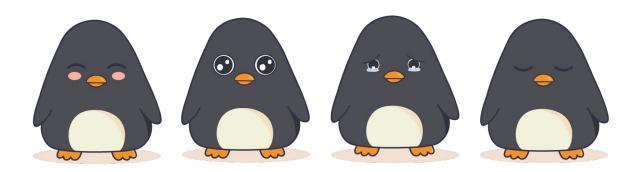
# Risks can be reasonably mitigated





Available NOW on Google Play!

# Thank you! Questions?







# Wob

# Anonymous geo-based messaging

David, Kyle, Kat, Jonathan, Julien, Sophia, Ehsan, Graham

# Appendix





### **Product Analysis**

### **Product Metrics**

### **Product**

- Num. of active users
- Num. of wobs
- Num. of comments

### **Business**

- Revenue
- Num. of sponsored posts
- Num. of new customers per month
- Average impressions per sponsored post

### **Technical**

- Uptime percentage
- Error rate



### **Competitor Analysis**

# Regulators

### Official Policing Bodies

 Highly aligned in keeping the online community a safe place

### Universities

- Give universities the right to moderate and delete content on campus grounds
- Aligned in keeping universities a welcoming place









# **Sponsors**

### Queen's University

- Mid to high alignment
- Better and more accessible platform to post campus and club activities





### **Local Restaurants**

- High alignment
- Community focused. Inexpensive platform to advertise on





### For Queen's University

# **Customer Acquisition Strategy**

### **Three Main Mediums**

### Digital



Sitting in Stauffer trying to procrastinate with the hundreds of other students around you? Resting in between sets at the ARC wondering what the gym drama is? Sitting in the lecture hall wondering if anybody else gets wtf is going on? Post and hear anonymously on Word on the Block

### **Experiential**



### Guerilla

Hit a PR? Want to shout out a friend making mad gains?
Want to elongate your rest?

Hear the Word on the Block

### For Queen's University

# **Customer Acquisition Strategy**

### **Three Main Mediums**

### **Digital**

### **Mediums:**

- Queen's Subreddit
- Facebook Groups
- Snapchat Geofilters

### **Experiential**

### Types:

- Laptop stickers
- Phone grips
- Normal stickers

### Guerilla

### **Mediums:**

- Location-Based Posters
- Queensu confessions

# **Pricing Model**

Year	1	2	3	4	5	6
Ads Revenue						
Users	15870	70095	173739	323690	489466	528956
New Users	18000	63480	126171	191112	226583	101962
Churn	2130	9255	22527	41161	60807	62472
Revenue	\$6,255.95	\$27,631.45	\$68,487.91	\$127,598.60	\$192,947.50	\$208,514.46
Company Ad Revenue						
Companies	7	35	86	161	244	264
Revenue	\$3,500.00	\$17,500.00	\$43,000.00	\$80,500.00	\$122,000.00	\$132,000.00
Total	\$9,755.95	\$45,131.45	\$111,487.91	\$208,098.60	\$314,947.50	\$340,514.46



# **Pricing Model**

Growth Rate Model						
Users	400%	130%	180%	110%	70%	45%
Retention	88.20%	89.00%	89.70%	90.40%	91.00%	91.50%
Churn	11.80%	11.00%	10.30%	9.60%	9.00%	8.50%

Assumptions	
Churn	10%
Activity Rate	20%
Cost per Click	0.27
Clicks per User	
per day	0.02
Days	365
Revenue per	
company	500



# **Churn Model**

alpha	1.53373				
beta	11.4273				
LL	-1644.71				
t	P(T=t)	S(t)	# active	# lost	
1	0.118334	0.881666	960	40	-85.3698
2	0.096858	0.784808	780	180	-420.212
3	0.080455	0.704353	653	127	-320.048
4	0.067683	0.63667	593	60	-161.575
5	0.057572	0.579098	551	42	-119.898
6	0.04945	0.529648	517	34	-102.231
7	0.042842	0.486806	491	26	-81.9059
					-353.466

### **Competitor Analysis**

# **Specific Competitor Analysis**





What Reddit gets Right

Good variety and separation between different types and categories of content What it doesn't

Hard to discover new content. Users generally stay within their subreddit bubble What Twitter gets Right

Low volume threads increase readability of its platform

What it doesn't

Content visibility is usually judged based on post history and number of followers

### MVP Requirements and WIP

## **Current Status & Timeline**

### **First** Second Semester Semester Ability to write WOB's Ability to write comments Map view and pin points Full-View version of WOR **Front End** Profile page and sign on Reporting system upvote/downvote Full server with database Neural network that uploaded to the internet determines trolling/abusive Back End Connecting backend to content frontend Operations to handle likes on comments

### **Moderation Strategy**

# Machine Learning & Downvoting

### Stage 1: Automatic Filter

Using a Neural Network in the backend, trained to identify unkind comments, it will scan requests for new WOBs before they are sent out to everyone, and if it determines that the new WOB to be unkind, with high confidence, that WOB will be prevented from release.

### Stage 2: Community Filter

WOBs and comments that receive a dramatic amount of downvotes, will be removed. Giving the opportunity to the WOB community to stand against cyberbullying and poor behaviour, just in case a more nuanced comment manage to pass the automatic filter.

#### Technical Architecture

# Implementing Automatic Filtering

- Gather dataset of troll comments in WOB, and/or use pre-existing datasets like Kaggle's twitter trolls dataset.
- Use the "BERT embedding layer" to map words/phrases into a Real Number
   Vector Space, as Machine Learning requires things to be modelled numerically
- Use the dataset to train Neural Networks with the goal to classify "troll or non troll comments". Trying a variety of different Deep Learning models such as "Convolutional Neural Networks", and "Recurrent Neural Networks", then evaluate their performance on a test dataset.
- Implement the best performing Neural Network into the backend, where new WOBs are received to be stored in the database, and inference the Neural Network to determine if it's a troll comment, and remove it if it determines it as such.



# Revenue

(Base Case)		Y1	Y2	Y3	Y4	Y5
Users		-	20,007	42,874	86,991	173,945
New Users		21,000	24,327	47,438	92,503	180,382
Churned Users		(993)	(1,460)	(3,321)	(5,550)	(10,823)
Total Users		20,007	42,874	86,991	173,945	343,503
Activity Rate %		90%	85%	80%	65%	50%
Yearly Active Users		18,006	36,443	69,593	113,064	171,752
Clicks per Active User		250	280	300	305	310
CPC	\$	0.01	\$ 0.01	\$ 0.02	\$ 0.02	\$ 0.02
Revenue		45,016	102,041	313,169	517,268	798,645