## FoodMeOnce



CS 373 Fall 2019 Group 9

## Meet the team!



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### About the Site

### **4** sprints

- Phase 1: Basic Static Website
- > Phase 2
  - Dynamic React Web Application
  - RESTful API
  - PostGres DB
- > Phase 3: Sorting, Searching, Filtering
- > Phase 4: Visualizations

### Models

- Districts
- Representatives
- Legislation

#### Links

- ➤ FoodMeOnce GitLab Repository
- ➤ FoodMeOnce API Documentation

## What is FoodMeOnce?

- Web Platform to allow users to gather information on food security throughout US
  Congressional Districts
- By combining Disparate Data sources we provide a well rounded perspective on food security in relation to political representation and legislation
- Generates Easy to understand Visualizations to allow a quick grasp of Congressional food support using various dimensions such as population, race, and representation.

### Database

state\_map

name text

short\_name\_text

state varchar(20)

congressional\_district

population varchar(7)

id integer

districts

varchar(2)

median\_age

gender\_ratio

mean\_income

varchar(6)

varchar(6)

poverty rate

varchar(6)

varchar(100)

num households

food\_insecurity\_rate

snap\_rate varchar(6)

state\_name varchar(2)

wiki\_page

integer

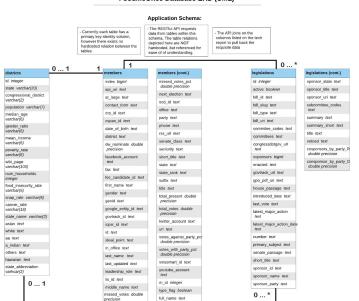
varchar(5)

cancer rate

varchar(18)

varchar(6)

#### FoodMeOnce Database ERD (UML)



#### Staging Schema:

 The python scripts dumped all the various data sources in different tables into this schema  From here, the application tables were made, joining data from the tables present in this schema

house_representatives	house_represental (cont.)
index integer	leadership_role fext
api_uri text	middle_name text
at_large boolean	missed_votes doub precision missed votes pct
contact_form text	
crp_id text	double precision
cspan_id text	next_election text
date_of_birth_text	ocd_id_text
district text	office text
dw_nominate double precision	party text
facebook account	phone text rss_url_text
text	
fax text	seniority text
fec_candidate_id text	short_title text
first_name_text	state text
gender text	suffix text
geoid text	title text
google_entity_id_text	total_present doub precision
govtrack_id text	
icpsr_id text	total_votes double precision
id text	twitter_account rex
ideal_point text	url text
in_office boolean	votes_against_part double precision
last_name text	
last_updated text	votes_with_party_p double precision
	votesmart_id text

youtube\_account

senators	sentitors (cont.)
index bigint	missed_votes bigint
api_uri text	missed_votes_pct double precision
contact_form text	
crp_id text	next_election text
cspan_id text	ocd_id_text
date_of_birth_text	office text
dw nominate double	party text
precision	phone text
facebook_account rext	rss_url_text
fax fext	senate_class text
	seniority text
fec_candidate_id_text	short_title_text
first_name_text	state fext
gender text	state_rank_text
google_entity_id_text	suffix text
govtrack_id text	title text
icpsr_id_text	total_present bigint
id text	total_votes bigint
ideal_point text	twitter_account fext
in_office boolean	url text
last_name_text	votes_against_party_pri double precision
last_updated text	
leadership_role_text	votes_with_party_pct double precision votesmart id text
lis_id_text	
niddle_name_text	_
	youtube_account text

legislations	legislations (cont.)
active boolean	sponsor_party text
bill_id_text	sponsor_state text
bill_slug_text	sponsor_title_text
bill_type_text	sponsor_uri text
bill_uri_text	subcomittee_codes
commitee_codes text	summary text
committees text	summary text
congressdotgov_url text	title text
coponsors bigint	vetoed text
enacted text	cosponsors_by_party_R double precision
govtrack_url_text	consponsor_by_party_E double precision
gpo_pdf_uri_text	
house_passage text	
introduced_date text	
last_vote text	
latest_major_action text	
latest_major_action_date rext	
number text	race_per_district
primary_subject_fext	state varchar(20)
senate_passage text	congressional_district varchar(20) race varchar(20) percentage double precision
short_title_text	
sponsor_id text	
sponsor_name text	
	state_name varchar(20)

## Tool stack

#### Front-end framework

- React Javascript
- Bootstrap and CSS
- **❖** Selenium
- **❖** Mocha
- **❖** D3

#### **Back-End Tools**

- PostgreSQL
- POSTMAN
- SQLAlchemy
- Flask
- Python

#### **IDEs**

- Pycharm IDE
- VSCode

#### **Backend**

❖ Amazon S3

#### Domain

- NameCheap
- Route53

#### **Others**

- Gitlab
- LucidChart
- Docker
- Gitlab CI/CD



## What did we do well?

- Made website dynamic in phase 1
- Searching, Sorting, Filtering
- Getting each phase reviewed by the TAs
- Well paced throughout each phase of the project.
- Gitlab Issue Board
- Communication with our Customer team

### What did we learn?

- ❖ E2E website building
  - Deploying and hosting in AWS
  - Designing an API
  - Domains and subdomains
  - Database design
- React and Flask frameworks
- Postman API documentation
- Importance of code readability

## What can we do better?

- ❖ UI/UX
  - Relative spacing of elements on screen
  - Did not consider mobile usage
  - Prettier visuals
  - Splash page
- Code refactoring/maintainability
- Could have used flask-restless
- District model page dynamic map load time

# What puzzled us?

- How to get started
- Asynchronous API requests
- Searching Data
- **❖** D3
- Gitlab CI setup

## Developer Team - PutltlnPark



## What did they do well?

- ❖ UI/UX is great
  - On mouse hovering over cards
  - Overall look and feel
  - Buttons
  - ➤ Visualizations located on model pages
  - Pagination has 5 pages
- Communicating on user stories was very thorough
- Made site mobile friendly

## What did we learn from their website?

#### Technical gains:

- Hovering effect over cards
- Using entire space for the instance page
- Use of icons(search icon, loading icon)

Website specific:

Central USA has the most frequency of parks + recreational activities

# What can they do better?

- Meeting phase requirements
  - > API subdomain
- ♦ Model page search results sometimes do not contain search term
  - National Parks search

## What puzzles us?

### Technical puzzles:

- How the model specific search works?
  - Possibly not correctly matching (ex) acadia in National Parks and tenn in States
  - No highlighting search terms on model specific search
- Website down at one point

Website specific:

Why does Central USA have such low visitor count despite having the most parks + recreational activities?

Questions?

Thank you!