

Documentation Recipe Macronutrients

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The healthiest diet: a collected dataset of recipes of various diets in terms of macronutrients

1. Motivation

1.1 For what purpose was the dataset created? Was there a specific task in mind? Was there a specific gap that needed to be filled? Please provide a description.

Consuming a healthy diet throughout a person's life helps prevent malnutrition in all its forms as well as a range of diet-related conditions and diseases (WHO, 2019). Recently, there are many types of diets available, all promising to offer the best and healthiest recipes to nourish a human body. While these recipes often include information on the nutritional value of these meals, the question remains whether sticking to these diets is really perceived as "healthy" and if using all the recommended recipes would actually provide your body with all the nutrients it needs.

There has been a lot of research and claims about which diet is healthier than the other. When you would break it down by nutrition, according to Cena and Calder (2020), a healthy diet consists of meals in which macronutrients (i.e. fats, proteins and carbohydrates) are consumed in the right proportions to support the body's physiological and energetic needs. Indeed, macronutrients provide the necessary energy for daily functioning. Additionally, micro nutrients (i.e., vitamins and minerals) are required in relatively small amounts for normal metabolism, development, growth and physiological functioning. However, the macronutrients are the three main sources of energy in food (Cena & Calder, 2020). Carbohydrates are mostly broken down in glucose and are the body's primary energy source. Fat aids in the integrity of cell membranes, the storage of energy, the protection of organs, the production of certain hormones, and the absorption of fat-soluble vitamins. The main role of protein is to maintain lean body mass, develop new tissues, and repair existing ones. In terms of macronutrients, numerous diet recommendations have been made to maintain health and prevent diseases (Flat, 2001). According to Ryan-Harshman & Aldoori (2006), most dietary references suggest that adults in general should consume 45% to 65% of their calories from carbohydrates, 20% to 35% from fat, and 10% to 35% from protein.

Because of the controversial opinion on which diet is perceived as healthy (and if certain diets are even healthy to follow) the aim of this project is to create a large dataset in which recipes of various diets are displayed with the main macronutrients: carbohydrates, fats and proteins. Researchers can use this dataset to study the effect of certain diets on the health of consumers. Furthermore, marketers can use the outcomes to discover patterns of macronutrients in the recipes of diets. This may help in improving the STP (Segmentation, Targeting, and Positioning) strategy of food manufacturing companies. Besides that, anyone who is interested in certain diets can extract information through the API about the recipe they are interested in.

When looking for a source to get our data from, we researched several options online. In doing so, we looked at different websites for scrapping, as well as other ways to get data, such as APIs and datasets. We first looked at a very popular food diary app called Myfitnesspal. We tried looking on their website to see if data

was publicly available, but we discovered that the data had to be obtained via an API, which is not free. We decided we needed to look wider to find the right data. We looked at other less popular food-tracking apps and websites, such as Fatsecret and Virtuagym, but we discovered that it is very difficult to scrape data from food-tracking apps, because the data is linked to personal data, and thus privacy issues can arise.

With this, we decided to include APIs and datasets and look at less popular food-tracking/nutritional websites from the Netherlands. Our idea was that here we would have a higher chance that the API would be publicly available. We found several websites, but some had too few recipes available or no nutritional information. Eventually, it led us to the website <https://www.edamam.com/>, which has a large food database and recipe API. We searched the website on Google Trends and found that the website is most popular in France and the US. We searched further and discovered that it is possible to get free access to the API (<https://developer.edamam.com/edamam-recipe-api>) as a developer. In our case, we thought it would be easier to use the API as it saves time and avoids legal problems. Edamam’s Recipe Search API is the endpoint used by this project. It is the only endpoint because it contains enough data that we want to collect. Table 1 gives a brief overview of the data sources, including the requirements we looked at.

Table 1: Overview of data sources.

Data source	Data extraction	Free access	Sufficient number of recipes available	Nutrition information available
https://www.taste.com/	Web scraping	Yes	No	Yes
https://www.themealdb.com/	API	Yes	No	No
https://zestfuldata.com/	API	Not for recipes	Yes	No
https://tasty.co/	Web scraping	Yes	Yes	Yes
https://www.edamam.com/	API	Yes	Yes	Yes
https://www.yummly.com/	API	No	Yes	Need premium account
https://www.bigoven.com/	Web scraping	Yes	Yes	Need premium account
https://www.fatsecret.com/	API	Yes	No	Yes, but only per product
https://www.nutritionix.com/	API	Yes	No	Yes, but only per product
https://www.hellofresh.com/	Web scraping	Yes	Hard to filter per diet	Yes

1.2 Who created this dataset (e.g., which team, research group) and on behalf of which entity (e.g., company, institution, organization)?

The dataset is created by team 12 from the course of Online Data Collection and Management of the master Marketing Analytics of Tilburg University. The team consist of the following members; Anouk Bor, Bi Xuan Guo, Mandana Khabbazi and Indi Wieggers. The instructor for the course and the project is Hannes Datta.

1.3 Who funded the creation of the dataset? If there is an associated grant, please provide the name of the grantor and the grant name and number.

This dataset was not created with funding from any organization or individual. Due to this fact, the data extraction code and datasets created by this team are entirely autonomous and made available to the public.

2. Composition

2.1 What do the instances that comprise the dataset represent (e.g., documents, photos, people, countries)? Are there multiple types of instances (e.g., movies, users, and ratings; people and interactions between them; nodes and edges)? Please provide a description.

Each instance in the dataset is a lunch/dinner recipe. These recipes are categorised into the five most popular diets for the users from <https://www.edamam.com/>. The users who use the Edamam site most frequently are located in Canada, Germany, France, the United Kingdom and India. The following diets are generally the most popular in the above countries (more on this in section 3.3); Keto diet, Mediterranean diet, Paleo diet, DASH diet and Vegan diet. The diets have multiple types of instances, including the macronutrients. These are categorized as fat, protein and carbohydrates (carbs). For each recipe, the macronutrients are listed in the database.

2.2 How many instances are there in total (of each type, if appropriate)?

Different numbers of recipes are available for all five diets. Table 2 shows a calculation of the total number of recipes available and the total number of recipes actually collected. The total number of instances in the dataframe is 34,898 in Edamam’s latest API. However, if the API changes, the number of instances can be altered. The API from Edamam has a limitation of a maximum of 10,000 calls per month and 10 calls per minute. One call from the Edamam API retrieves 20 recipes. Therefore, it takes $34,898/20 = 1744.9$ requests to retrieve the data of all five diets. Due to the limitation of the API, it takes $1744.9/10/60 = 2.9$ hours to gather all data. Since the number of requests required does not exceed the limit of 10,000 calls per month, the data can be easily collected with a single API key. Although our goal was to retrieve all available instances, our team encountered technical difficulties in scaling up the data collection, which are detailed in section 3.1. Thus, the actual number of instances collected differs from what is available, as shown in Table 2.

Table 2: Number of instances available and actually collected.

Diet type	Number of recipes available	Number of recipes collected
DASH	10000	1745
Keto	2323	1512
Mediterranean	4635	1753
Paleo	7940	1274
Vegan	10000	1522
Total	34898	7806

2.3 Does the dataset contain all possible instances or is it a sample (not necessarily random) of instances from a larger set? If the dataset is a sample, then what is the larger set? Is the sample representative of the larger set (e.g., geographic coverage)? If so, please describe how this representativeness was validated/verified. If it is not representative of the larger set, please describe why not (e.g., to cover a more diverse range of instances, because instances were withheld or unavailable).

This dataset is a sample of the entire Edamam recipe API. The Recipe Search API contains over 2.3 million recipes. As mentioned earlier, the dataset only contains recipe information from the diets we are interested in. An attempt was made to collect all recipes from the five diets so that the sample represents each diet separately. However, a total of 7,806 instances were collected out of a total of 34,898 possible instances. Therefore, the sample may not be representative of the larger set, because instances are no longer collected when scaling up the data collection.

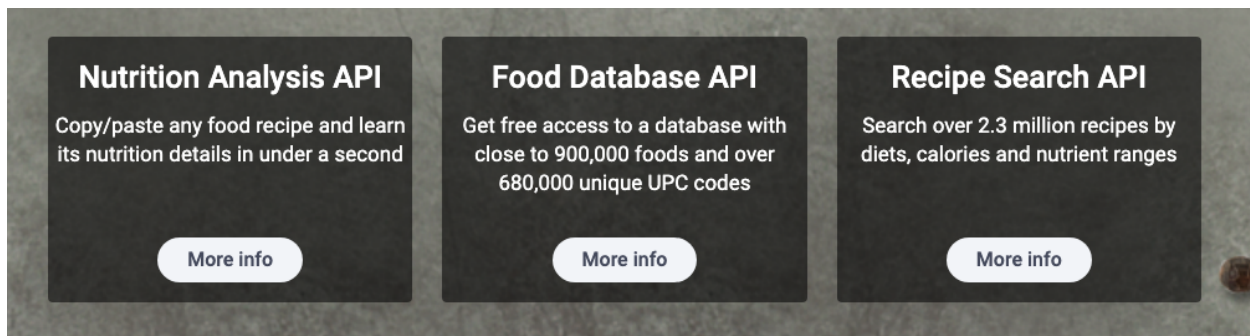


Figure 1: Edamam’s APIs.

2.4 What data does each instance consist of? “Raw” data (e.g., unprocessed text or images) or features? In either case, please provide a description.

For the raw data, 7,806 instances were collected. The raw data is stored in a .json output for further processing. Table 3 contains the descriptive statistics of the five diets.

Table 3: Descriptive statistics macronutrients per diet.

Diet type	Mean carbs	SD carbs	Mean fats	SD fats	Mean protein	SD protein
DASH	161	206	101	134	69.3	91.5
Keto	58	78.1	153	116	101	98.9
Mediterranean	153	165	101	97	101	93.9
Paleo	130	161	136	135	88.7	97.8
Vegan	254	223	103	120	56.2	47.2

The code creates a dataset containing the health label, meal type, cuisine type and corresponding macronutrients for each recipe. The macronutrients are given in integer variables, allowing users to calculate with the numbers. The following tables explicitly list the variables used in the code.

Table 4: Nutrient structure.

Variable	Data type	Description
uri	String	Ontology identifier
label	String	Display label
quantity	Float	Quantity of specified units
unit	String	Units

Table 5: Ingredient structure.

Variable	Data type	Description
foodId	String	Food identifier
quantity	Float	Quantity of specified measure
measure	Measure	Measure
weight	Float	Total weight in grams
food	Food	Food
foodCategory	String	Shopping aisle category

Table 6: Health labels.

Type	Web label	API Parameter	Definition
Health	Alcohol-Cocktail	alcohol-cocktail	Describes an alcoholic cocktail
Health	Alcohol-Free	alcohol-free	No alcohol used or contained
Health	Celery-Free	celery-free	Does not contain celery or derivatives
Health	Crustcean-Free	crustacean-free	Does not contain crustaceans (shrimp, lobster etc.) or derivatives
Health	Dairy-Free	dairy-free	No dairy; no lactose
Health	DASH	DASH	Dietary Approaches to Stop Hypertension diet
Health	Egg-Free	egg-free	No eggs or products containing eggs
Health	Fish-Free	fish-free	No fish or fish derivatives
Health	FODMAP-Free	fodmap-free	Does not contain FODMAP foods
Health	Gluten-Free	gluten-free	No ingredients containing gluten
Health	Immuno-Supportive	immuno-supportive	Recipes which fit a science-based approach to eating to strengthen the immune system
Health	Keto-Friendly	keto-friendly	Maximum 7 grams of net carbs per serving
Health	Kidney-Friendly	kidney-friendly	Per serving – phosphorus less than 250 mg AND potassium less than 500 mg AND sodium less than 500 mg
Health	Kosher	kosher	Contains only ingredients allowed by the kosher diet. However it does not guarantee kosher preparation of the ingredients themselves
Health	Low Potassium	low-potassium	Less than 150mg per serving
Health	Low Sugar	low-sugar	No simple sugars – glucose, dextrose, galactose, fructose, sucrose, lactose, maltose
Health	Lupine-Free	lupine-free	Does not contain lupine or derivatives
Health	Mediterranean	Mediterranean	Mediterranean diet
Health	Mollusk-Free	mollusk-free	No mollusks
Health	Mustard-Free	mustard-free	Does not contain mustard or derivatives
Health	No oil added	No-oil-added	No oil added except to what is contained in the basic ingredients
Health	Paleo	paleo	Excludes what are perceived to be agricultural products; grains, legumes, dairy products, potatoes, refined salt, refined sugar, and processed oils
Health	Peanut-Free	peanut-free	No peanuts or products containing peanuts
Health	Pescatarian	pecatarian	Does not contain meat or meat based products, can contain dairy and fish
Health	Pork-Free	pork-free	Does not contain pork or derivatives
Health	Red-Meat-Free	red-meat-free	Does not contain beef, lamb, pork, duck, goose, game, horse, and other types of red meat or products containing red meat.
Health	Sesame-Free	sesame-free	Does not contain sesame seed or derivatives
Health	Shellfish-Free	shellfish-free	No shellfish or shellfish derivatives
Health	Soy-Free	soy-free	No soy or products containing soy
Health	Sugar-Conscious	sugar-conscious	Less than 4g of sugar per serving
Health	Sulfite-Free	sulfite-free	No Sulfites
Health	Tree-Nut-Free	tree-nut-free	No tree nuts or products containing tree nuts
Health	Vegan	vegan	No meat, poultry, fish, dairy, eggs or honey
Health	Vegetarian	vegetarian	No meat, poultry, or fish
Health	Wheat-Free	wheat-free	No wheat, can have gluten though

Table 7: Meal types.

Type	Value
mealType	breakfast
mealType	brunch
mealType	lunch/dinner
mealType	snack
mealType	teatime

Table 8: Cuisine types.

Type	Value
cuisineType	american
cuisineType	asian
cuisineType	british
cuisineType	caribbean
cuisineType	central europe
cuisineType	chinese
cuisineType	eastern europe
cuisineType	french
cuisineType	greek
cuisineType	indian
cuisineType	italian
cuisineType	japanese
cuisineType	korean
cuisineType	kosher
cuisineType	mediterranean
cuisineType	mexican
cuisineType	middle eastern
cuisineType	nordic
cuisineType	south american
cuisineType	south east asian
cuisineType	world

2.5 Is there a label or target associated with each instance? If so, please provide a description.

Each instance, in this case each recipe, is labeled under *hits* and *recipe*. This is the same for each dataset per diet.

2.6 Is any information missing from individual instances? If so, please provide a description, explaining why this information is missing (e.g., because it was unavailable). This does not include intentionally removed information, but might include, e.g., redacted text.

There is no information missing from individual instances, however, the number of calories in some recipes seems to be wrong. The number seems oddly high (see Figure 2 for an example). Additionally, the data does not include a serving size, which could diminish the number of calories per serving. Thus, after checking the original source of the recipes, it can be assumed that the information is not accurate. Therefore, calorie counts are removed when creating a dataset from the API. This is to avoid spreading inaccurate information about recipes, which may lead to wrong conclusions.

```

"totalNutrients": {
  "ENERC_KCAL": {
    "label": "Energy",
    "quantity": 3898.1978808751583,
    "unit": "kcal"
  }
}

```

Figure 2: Example of the calories.

2.7 Are relationships between individual instances made explicit (e.g., users' movie ratings, social network links)? If so, please describe how these relationships are made explicit.

The relationships between individual instances are made explicit by the link of the recipe to the original site from which the API retrieved the information. These instances are identified under *source* and the link to the website under *url*.

2.8 Are there any recommended data splits (e.g., training, development/validation, testing)? If so, please provide a description of these splits, explaining the rationale behind them.

The .json file contains all the collected raw data. A recommended data split could be based on the amount of macronutrients per recipe based on specific diets that users of the dataset follow. Another split could be based on certain types of cuisines preferred by users. Other splits are also possible, but the above splits are recommended by the dataset creators, taking into account the purpose of this dataset.

2.9 Is the dataset self-contained, or does it link to or otherwise rely on external resources (e.g., websites, tweets, other datasets)? If it links to or relies on external resources, a) are there guarantees that they will exist, and remain constant, over time; b) are there official archival versions of the complete dataset (i.e., including the external resources as they existed at the time the dataset was created); c) are there any restrictions (e.g., licenses, fees) associated with any of the external resources that might apply to a future user? Please provide descriptions of all external resources and any restrictions associated with them, as well as links or other access points, as appropriate.

Each instance in the Edamam API is linked to an external resource, namely a website URL. There is no guarantee that the external websites will persist over time, as Edamam reserves the right to update, modify or alter the Edamam API. However, Edamam will endeavour to announce such updates, modifications, or alterations in advance which will also allow the dataset to undergo necessary updates, changes or modifications. And even if an external source is removed, no data is lost from the API, as the dataset is entirely dependent on the Edamam API. As of October 16, 2022, there are no restrictions (e.g. licences, fees) on the external resources that could apply to a future user.

2.10 Does the dataset contain data that might be considered confidential (e.g., data that is protected by legal privilege or by doctor-patient confidentiality, data that includes the content of individuals non-public communications)? If so, please provide a description.

This dataset does not contain data that one could consider confidential. This is based on the fact that the data is publicly available on the Edamam API website. There is no information in the data that could trace

back to private information (personal data, locations etc.).

2.11 Does the dataset contain data that, if viewed directly, might be offensive, insulting, threatening, or might otherwise cause anxiety? If so, please describe why.

The dataset does not contain data that could be considered offensive, insulting, or threatening. The only thing that the dataset contains is information about precise food measures, which might trigger feelings for people with an eating disorder. There is a possibility that the information can cause feelings of anxiety for this group of people.

2.12 Does the dataset relate to people? If not, you may skip the remaining questions in this section.

The dataset does not relate to people. Therefore, questions 2.12 to 2.15 were not answered and deleted from the documentation.

3. Collection Process

3.1 How was the data associated with each instance acquired? Was the data directly observable (e.g., raw text, movie ratings), reported by subjects (e.g., survey responses), or indirectly inferred/derived from other data (e.g., part-of-speech tags, model-based guesses for age or language)? If data was reported by subjects or indirectly inferred/derived from other data, was the data validated/verified? If so, please describe how.

The data of each instance is directly observable as raw text and accessed using Edamam’s API. The entire collection process consists of several steps and is stored in a script with written code in ‘src/collection/database_recipe_API.py’ and ‘src/collection/database_recipe_API.ipynb’.

First, the script creates a .json file that collects all the raw data for a diet. This can be done for any diet, and in this case, five separate .json files are created for the five diets. Next, all the recipes for a diet are extracted using a while loop, ensuring that the code gets 6 seconds of sleep due to the limitation of 10 calls per minute mentioned in section 2.2. After this, the code extracts the macronutrients from all recipes. Again, the sleep time is set to 6 seconds because of the limitation. In addition, the script creates a separate CSV file to have a clear overview of the macronutrient information for each recipe. Once more, separate files are created for each diet, creating a total of five CSV files to see which recipe belongs to which diet. To put everything together, the code merges the different CSV files into one big CSV file with information from all five diets. Finally, the “pandas” package was used to structure the data into a large table (dataframe) with all variables of interest.

Note that the script can specify which diets the data is scraped from. The number of diets scraped therefore depends on what the user of this scraper is interested in, so the number of CSV files generated may also vary.

Unfortunately, our team encountered technical problems when scaling up the data collection. For each diet, an error appeared for a particular line when trying to extract more data. An example of the error is shown in Figure 3. The debugging phase consisted of looking up the error online, whether other users of this API had the same problems, and trying to parse these specific lines in the .json files. However, we could not figure out what caused this error. Our guess is that it has to do with the output of the .json files becoming too large. Although we failed to collect data from all instances, our dataset is still large enough and contains enough information to use.


```

Getting data for 1940
Getting data for 1960
Getting data for 1980
7819
headersize: None
<!doctype html><html lang="en"><head><title>HTTP Status 400 – Bad Request</title><style type="text/css">body {font-family:Tahoma,Arial,sans-serif;} h1, h2, h3, b {color:white;background-color:#525D76;} h1 {font-size:22px;} h2 {font-size:16px;} h3 {font-size:14px;} p {font-size:12px;} a {color:black;} .line {height:1px;background-color:#525D76;border:none;}</style></head><body><h1>HTTP Status 400 – Bad Request</h1><hr class="line" /><p><b>Type</b> Exception Report</p><p><b>Message</b> Request header is too large</p><p><b>Description</b> The server cannot or will not process the request due to something that is perceived to be a client error (e.g., malformed request syntax, invalid request message framing, or deceptive request routing).</p><p><b>Exception</b></p><pre>java.lang.IllegalArgumentException: Request header is too large
    org.apache.coyote.http11.Http11InputBuffer.fill(Http11InputBuffer.java:779)
    org.apache.coyote.http11.Http11InputBuffer.parseHeader(Http11InputBuffer.java:946)
    org.apache.coyote.http11.Http11InputBuffer.parseHeaders(Http11InputBuffer.java:593)
    org.apache.coyote.http11.Http11Processor.service(Http11Processor.java:284)
    org.apache.coyote.AbstractProcessorLight.process(AbstractProcessorLight.java:65)
    org.apache.coyote.AbstractProtocol$ConnectionHandler.process(AbstractProtocol.java:893)
    org.apache.tomcat.util.net.NioEndpoint$SocketProcessor.doRun(NioEndpoint.java:1707)

```

Figure 3: Example of the error in scaling up data collection.

3.2 What mechanisms or procedures were used to collect the data (e.g., hardware apparatus or sensor, manual human curation, software program, software API)? How were these mechanisms or procedures validated?

As mentioned in the data collection, the API of Edamam was used to obtain the data of the different recipes of the diets. In order to validate this data, the different recipes and their labels and macronutrients were compared with the recipe on the site of Edamam. We also compared the recipe with the original site offered within the API.

We scraped the data using the web application Jupyter Notebooks launched by Anaconda Navigator. Loading the packages requests, json, pandas, time and os allowed us to use functions that enabled our data collection steps.

3.3 If the dataset is a sample from a larger set, what was the sampling strategy (e.g., deterministic, probabilistic with specific sampling probabilities)?

The dataset is a sample from a larger set obtained through deterministic sampling. Based on the site of PlanKetogenic (n.d.) that contains the most researched diets worldwide, the most popular diets in the countries that use Edamam the most were chosen and filtered for the dataset. For each diet, the dataset was filtered for the macronutrients fats, carbohydrates and proteins to see how many grams each recipe contains. The data set was also filtered by lunch/dinner. The reason behind this is because intermediate fasting (where someone consumes their food between 12pm and 8am) has become very popular in recent years (Mohiuddin, 2018; Snyder, 2022). This means that people more often choose to skip breakfast, so lunch/dinner seemed a more useful meal type to extract information from. The cuisine type was also added to see which cuisine the recipe came from. The dataset was filtered for the first 8000 recipes, as the software programs used (Jupyter Notebook and Anaconda) could not otherwise run the script due to licensing issues. As a result, it is unfortunately not possible to automate the script. However, the scraper does comment on what one should do to get the CSV file that aggregates the nutritional information of all diets. This also makes it easier for future users of the scraper, as there is a guide on how to add the diets to the scraper one is interested in.

3.4 Who was involved in the data collection process (e.g., students, crowdworkers, contractors) and how were they compensated (e.g., how much were crowdworkers paid)?

Only the students of this project were involved in the data collection process. The instructor of the project and course did provide assistance during data collection. There was no compensation as this was not applicable to this project.

3.5 Over what timeframe was the data collected? Does this timeframe match the creation timeframe of the data associated with the instances (e.g., recent crawl of old news articles)? If not, please describe the timeframe in which the data associated with the instances was created.

The data collected for the project were collected in real time. They were collected in real time because they are available immediately after collection. Time stamps were added to represent the day and time when the data were collected. The final dataset was collected on the 16th of October and is stored in the file "All_Diets.csv".

3.6 Were any ethical review processes conducted (e.g., by an institutional review board)? If so, please provide a description of these review processes, including the outcomes, as well as a link or other access point to any supporting documentation.

There were no ethical review processes conducted. Therefore, this question is not applicable.

3.7 Does the dataset relate to people? If not, you may skip the remaining questions in this section.

The dataset does not relate to people, because the Edamam API does not contain any user-specific information whatsoever. Therefore, the questions in the following section are not applicable.

4. Preprocessing, cleaning, labeling

4.1 Was any preprocessing/cleaning/labeling of the data done (e.g., discretization or bucketing, tokenization, part-of-speech tagging, SIFT feature extraction, removal of instances, processing of missing values)? If so, please provide a description. If not, you may skip the remainder of the questions in this section.

As mentioned, raw data was gathered using the API and stored in CSV files. To obtain the final dataset, we carried out several preprocessing steps. First, it is important to note that tokenisation must be used for the API key. The API key contains sensitive data that must be converted into non-sensitive data, called "tokens", to be used in the script without exposing the sensitive information. This was done by configuring environment variables.

There were no missing values in the raw data, so no missing values were removed. As previously explained, calorie counts did not seem reliable, hence they were not included in the dataset. Micronutrients were not added to the dataset as it focuses on macronutrients. Labels were created for diet type, recipe name, cuisine type and the amounts of proteins, carbohydrates and fats in grams, respectively. The final dataset consists of merging the different CSV files of the diets, which is stored in the file 'All_Diets.csv'.

4.2 Was the "raw" data saved in addition to the preprocessed/cleaned/labeled data (e.g., to support unanticipated future uses)? If so, please provide a link or other access point to the "raw" data.

As aforementioned, the raw data for each diet is automatically stored in .json files, alongside the preprocessed/cleaned/labeled data as part of our coding script. This is also particularly useful to support unexpected future applications. The raw data for each diet is stored in a file named '[diet]_raw_data.json'.

4.3 Is the software used to preprocess/clean/label the instances available? If so, please provide a link or other access point.

For tokenisation and configuring environment variables, the steps were followed in the following link: <https://tilburgsciencehub.com/building-blocks/store-and-document-your-data/store-data/environment-variables/>. Mac/Linux users should use terminal and accessing the environment variables is possible with Python or R. In addition, Anaconda, Jupiter Notebook and the packages needed for functions to preprocess/clean/label the data are all publicly available.

5. Uses

5.1 Has the dataset been used for any tasks already? If so, please provide a description.

The data set has not yet been used for any tasks, but is available for future use.

5.2 Is there a repository that links to any or all papers or systems that use the dataset? If so, please provide a link or other access point.

There is no repository available yet for papers and systems that use the data scraped in this project. However, we have our repository that contains all the files and the final dataset corresponding to this project. The link to this repository is: <https://github.com/Mandanakhabbazi/recipe-nutritions.git>

5.3 What (other) tasks could the dataset be used for?

Researchers can use this dataset to study the effect of the five diets on consumer health. Furthermore, marketers can use the outcomes to discover patterns of macronutrients in the diets' recipes. This can help improve food producers' STP (Segmentation, Targeting, and Positioning) strategy. Although this dataset focuses only on the diets Keto, vegan, DASH, Mediterranean and Paleo, one can in fact create datasets of any diet one is interested in, as the script allows to define the diet one wants to scrape and is not limited to the five diets used in this project. This allows researchers and marketers to explore other diets as well.

5.4 Is there anything about the composition of the dataset or the way it was collected and preprocessed/cleaned/labeled that might impact future uses? For example, is there anything that a future user might need to know to avoid uses that could result in unfair treatment of individuals or groups (e.g., stereotyping, quality of service issues) or other undesirable harms (e.g., financial harms, legal risks) If so, please provide a description. Is there anything a future user could do to mitigate these undesirable harms?

How the data is pre-processed is crucial. As mentioned, confidential information should be tokenised by configuring environment variables. As far as we know, there is nothing else about the composition of the dataset or how it was collected and pre-processed/cleaned/labelled that could affect future use.

5.5 Are there tasks for which the dataset should not be used? If so, please provide a description.

While the data set can be used as a dietary guide, it should not be used as a sole guideline, as it focuses only on the macro-nutrients carbohydrates, fats and proteins. Moreover, consumers cannot tell from our dataset how many calories are in a serving of a recipe. In addition, the amounts of vitamins and minerals are also not listed with the recipes, even though they contribute a lot to a person's health.

The dataset is provided using the free version of the Edamam API and is intended for personal or non-profit use. Thereafter, it is prohibited to use the free version of the Edamam API in connection with any

commercial or business purpose or in any manner that is commercial or business oriented without the express permission of Edamam.

6. Sources

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