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# Task 1

## Checkpoint 1

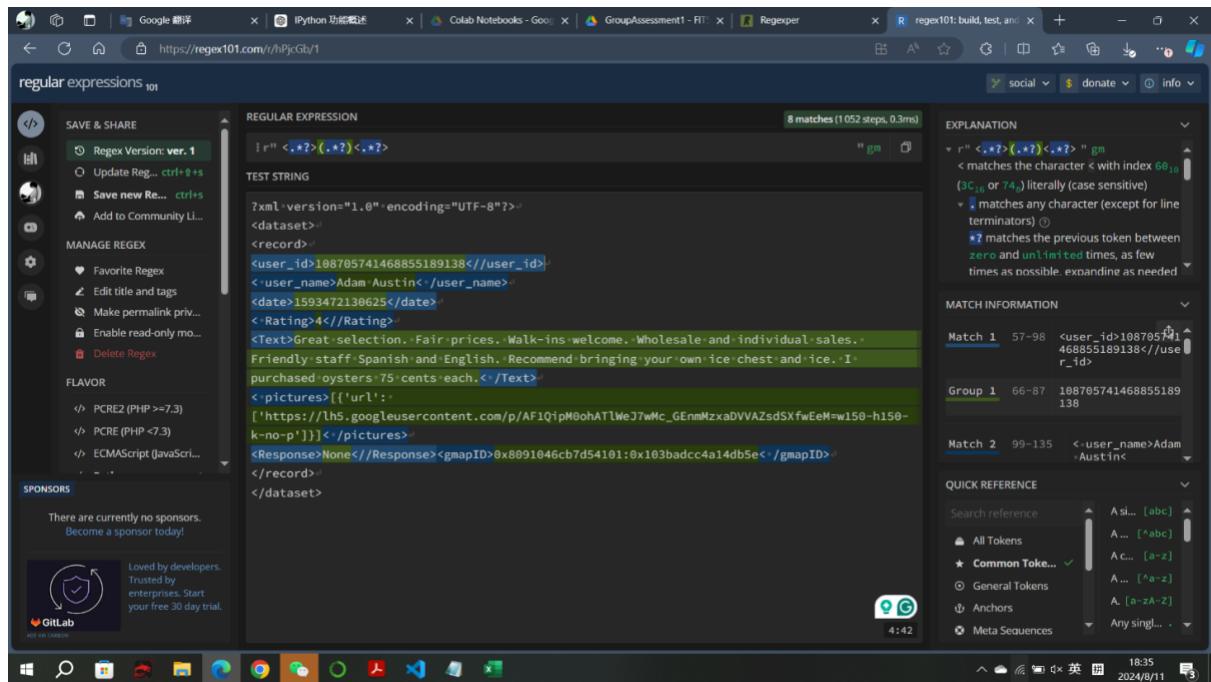
**DateTime:** 2024-08-11 18:35

**Content:** Try to use regex to extract the .txt data.

**Task 1 State:** task1\_020.ipynb can load the .xlsx file.

**Contribution:** Zihan Yin

**Proof:**



## Checkpoint 2

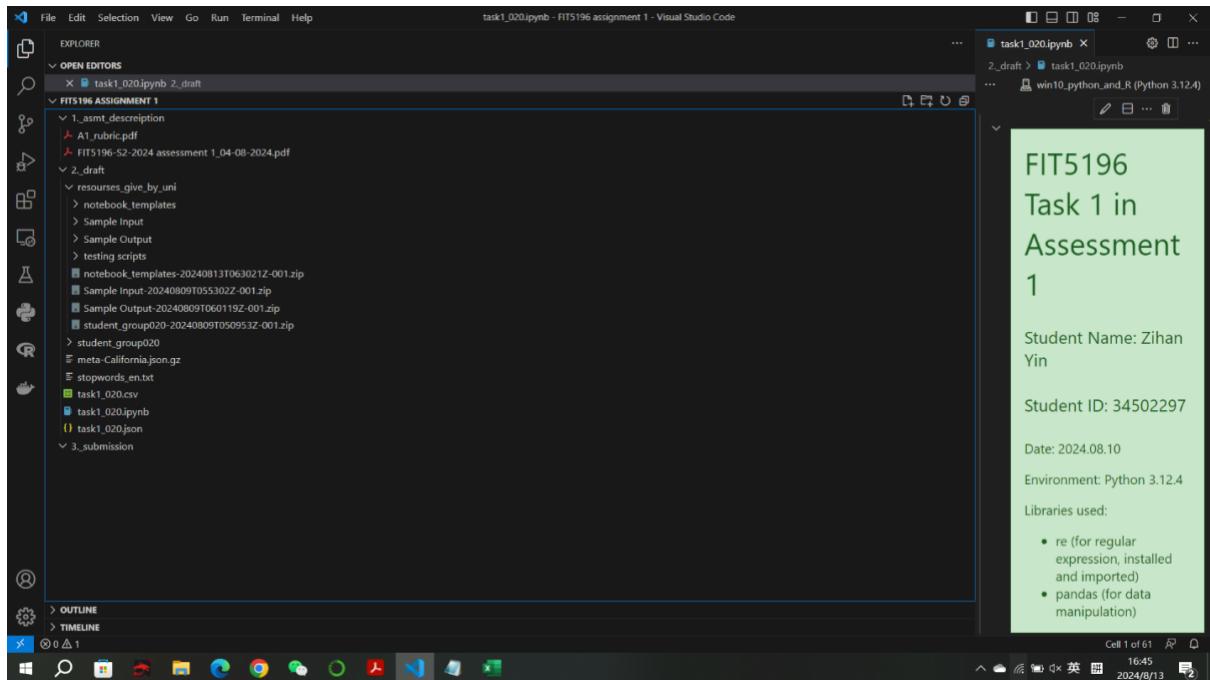
**DateTime:** 2024-08-13 16:45

**Content:** Finish all the codes of task 1.

**Task 1 State:** task1\_020.ipynb can output the required .csv file & .json file, which are able to pass task1\_test.py

**Contribution:** Zihan Yin

## Proof:



## Checkpoint 3

**Date****Time:** 2024-08-16 4:28

### Content:

Complete code comments, step explanations, and framework structures for task 1.

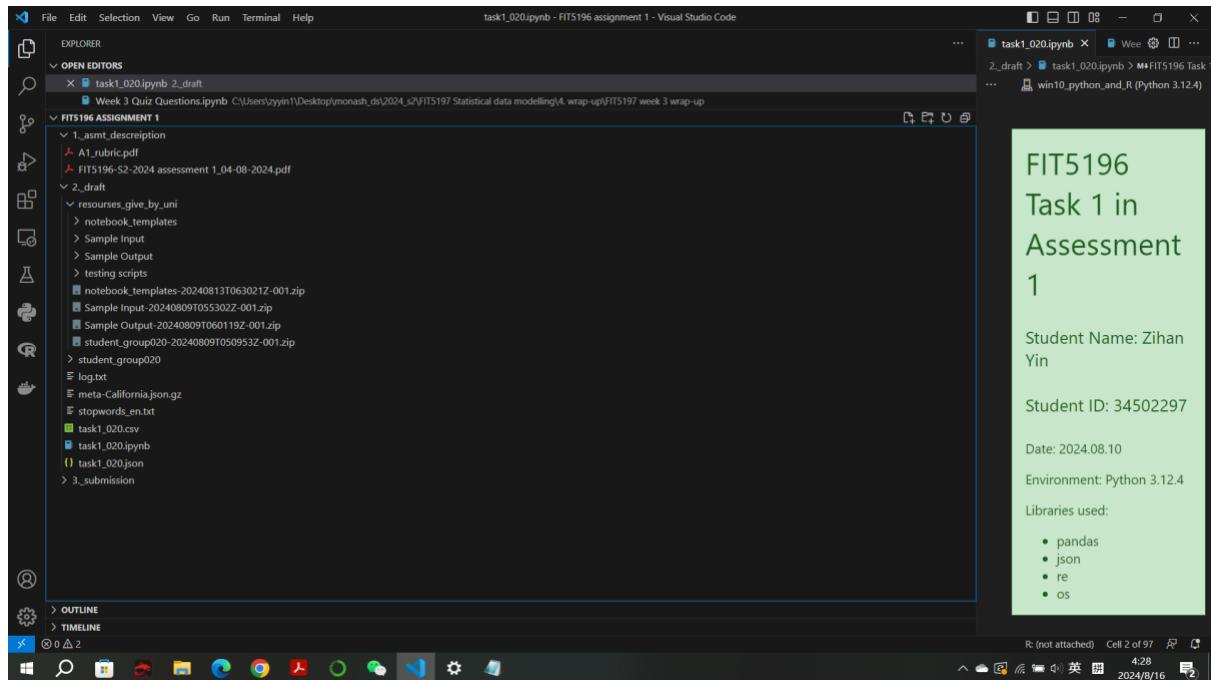
In particular:

1. Adjust the markdown format and write the interpretation
2. Adjust the code to reduce redundancy
3. Add appropriate code to show counterexamples
4. Write code comments
5. Check the accuracy repeatedly

**State:** all done with better readability.

**Contribution:** Zihan Yin

## Proof:



## Task 2

### Checkpoint 1

**Date/Time:** 2024-08-24 18:12

**Content:** Extract meaningful bigrams.

**Challenge:** The output of bigrams after stemming does not meanigful anymore, words becomes like 'earli\_smell' which is hard to be analysis. For words to be meaningful, the only way is to keep their original form. Thus, it is decided to not stem that bigrams. The challenge here is that how to stem unigram but not stem bigrams? The solution I apply here is to create two empty lists called 'all\_tokens' and 'unigram\_list'. One stores the token that only has been removed stop words from 'stopwords\_en.text', while another one stores tokens that has been stemmed, removed stops as well as other threshold. In this case, bigrams and unigrams are operated in differnet lists. After finishing all the cleaning step, they are combined together and stored in a list called 'final\_vocabulary'.

**Contribution:** Ruiwen Chen

## Proof:

```
['ach_dress', 'approv_sanit', 'aren_guidelin', 'attack_pound', 'bank_account', 'bare_cloth', 'broom_closet', 'chill_motiv', 'closet_ton', 'code_tire', 'common_courtesi', 'credit_bank', 'diego_angel', 'dress_code', 'drove_simpli', 'earlier_sat', 'effect_compar', 'excus_inexpens', 'expert_excus', 'extend_period', 'gain_wanna', 'gentleman_drove', 'gripe_true', 'guidelin_overnight', 'haha_mark', 'head_spray', 'healthier_activ', 'heart_attack', 'hire_switch', 'h_oop_easier', 'ignor_request', 'impecc_million', 'improv_health', 'info_file', 'jump_hoop', 'larger_internet', 'main_tain_van', 'march_mile', 'million_longer', 'motiv_crew', 'nation_begin', 'neg_obviou', 'news_gold', 'opinion_hire', 'overnight_clear', 'period_mad', 'potenti_skin', 'pound_tip', 'scale_greatest', 'scan_sight', 'shame_are', 'shell_larger', 'sight_unsur', 'skin_damag', 'sleep_strip', 'string_bare', 'suck_sleep', 'summer_support', 'sunday_fat', 'switch_system', 'tire_string', 'vacat_tub', 'van_approv', 'video_maintain', 'wall_freak', 'activ_rule', 'age_amen', 'alot_scale', 'angel_san', 'anxieti_march', 'averag_pictur', 'beginn_cash', 'bodi_enhanc', 'cash_credit', 'club_out_dat', 'condit_hassl', 'countri_boyfriend', 'damag_age', 'dim_guest', 'earli_saturday', 'easiest_transact', 'fact_ju dg', 'freak_boyfriend', 'health_judg', 'inexpens_smell', 'intimid_shape', 'judg_nervou', 'kick_weekend', 'kinda_stu pid', 'kinda_weird', 'light_dim', 'light_suck', 'longer_weekend', 'lose_info', 'main_expert', 'meet_class', 'monito r_club', 'muscl_kinda', 'nose_possibl', 'obviou_broken', 'pictur_muscl', 'posit_skip', 'possibl_news', 'privaci_wom en', 'privat_stall', 'pull_nose', 'pull_shame', 'quit_anxieti', 'rule_improv', 'san_diego', 'sat_possible', 'saturda y_sunday', 'season_bodi', 'shape_differ', 'smell_sour', 'sour_stale', 'spray_wide', 'stale_wet', 'stand_meet', 'str ip_light', 'stupid_pictur', 'system_wifi', 'thursday_understaf', 'tip_lose', 'tortilla_awsom', 'tortilla_key', 'tra iner_unlimit', 'true_saturday', 'understaf_thursday', 'unlimit_train', 'unprofession_lose', 'vega_san', 'wanna_priv at', 'weekend_earli', 'weekend_earlier', 'wet_tortilla', 'wide_singl', 'wifi_shell', 'wifi_video', 'worn_pull', 'ac
```

```
['accommodating_providing', 'aches_pains', 'active_lifestyle', 'age_soledad', 'annoying_manning', 'answers_question s', 'appeared_broom', 'aren_guidelines', 'attacks_stroke', 'automated_system', 'bandwidth_internet', 'bank_account' , 'bar_lights', 'barely_clothes', 'beginner_short', 'beginning_exerciser', 'broom_closet', 'changing_stalls', 'char ged_contract', 'childcare_class', 'classes_pools', 'club_outdated', 'code_tired', 'common_courtesy', 'completely_sh ame', 'cook_aches', 'damage_premature', 'daunting_task', 'didn_single', 'diego_los', 'dress_code', 'dumbbells_lbs', 'earlier_sat', 'early_saturday', 'effective_comparing', 'elliptical_rowing', 'enforcevtgr_dress', 'equipped_wall', 'exercising_public', 'extended_periods', 'facilities_late', 'fact_judge', 'family_arthritis', 'fart_cook', 'fast_up load', 'fees_haven', 'fox_news', 'francisco_denver', 'gains_end', 'garden_hose', 'giving_prayer', 'greatest_accommo dating', 'gripe_true', 'grocery_store', 'groups_monopolizing', 'grunting_rule', 'haha_thought', 'hassles_judgement' , 'head_rusted', 'healthier_active', 'heart_attacks', 'heavier_dumbbells', 'hose_equipped', 'ignore_request', 'imme diately_freaked', 'include_privacy', 'info_letting', 'internet_busy', 'intervention_acknowledging', 'jump_hoops', 'lap_tool', 'large_number', 'larger_bandwidth', 'late_night', 'lb_pounds', 'left_travel', 'letting_piling', 'license_year', 'lifestyle_grunting', 'lighting_sucks', 'lights_dim', 'los_angeles', 'lukewarm_partner', 'lunk_rules', 'main tained_grocery', 'man_improve', 'manager_enforce', 'mark_ps', 'masks_refuse', 'massages_worst', 'meet_classes', 'me mberships_technically', 'miles_town', 'minded_giving', 'morning_favorite', 'moothly_provide', 'nation_beginning', 'n eeding_cleaned', 'negatives_obvious', 'nervous_exercising', 'ny_vegas', 'office_playing', 'online_daunting', 'op_fa rt', 'order_meet', 'overnight_clears', 'overweight_man', 'pains_enforcevtgr', 'pf_membejuries', 'pictures_cartoon' , 'piling_fees', 'pleasant_understanding', 'podunk_towns', 'policy_signs', 'pools_organized', 'potential_skin', 'pr ayer_drove', 'preening_posing', 'premature_aging', 'product_excuse', 'professional_fat', 'provide_bank', 'radius_di
```

## Checkpoint 2

**Date**: 2024-08-25 16:00

**Challenge:** Not sure how to check whether the 200 bigrams are in the text. Initially, I am thinking to use a loop to loop through both bigrams and text, if they match, they append in a list. This one works, but it might be less effective because it needs to loop a lot. Then, I am thinking if `ngram` works or not, it turns out it is the same output as the loop one, then, I choose to with this because it is more concise.

**Contribution:** Ruiwen Chen

## Proof:

```
valid_bigrams = []
for bigram in bigram_vocab:
    word1, word2 = bigram.split('_')

    # Check if word1 and word2 appear consecutively in the original tokenized text
    for i in range(len(all_tokens) - 1):
        if all_tokens[i] == word1 and all_tokens[i + 1] == word2:
            valid_bigrams.append(bigram)
            break # Once found, no need to check further for this bigram

print(valid_bigrams)
print(len(valid_bigrams))

['accommodating_providing', 'aches_pains', 'active_lifestyle', 'annoying_manning', 'answers_questions', 'appeared_broom', 'aren_guidelines', 'automated_system', 'bank_account', 'barely_clothes', 'beginning_exerciser', 'broom_close_t', 'changing_stalls', 'charged_contract', 'childcare_class', 'classes_pools', 'club_outdated', 'code_tired', 'comm_on_courtesy', 'completely_shame', 'damage_premature', 'daunting_task', 'diego_los', 'dress_code', 'dumbbells_lbs', 'earlier_sat', 'early_saturday', 'effective_comparing', 'elliptical_rowing', 'equipped_wall', 'exercising_public', 'extended_periods', 'fact_judge', 'family_arthritis', 'fees_haven', 'greatest_accommodating', 'gripe_true', 'groups_monopolizing', 'grunting_rule', 'hassles_judgement', 'head_rusted', 'healthier_active', 'heart_attacks', 'ignore_rqquest', 'immediately_freaked', 'include_privacy', 'info_letting', 'intervention_acknowledging', 'jump_hoops', 'letting_piling', 'lifestyle_grunting', 'lighting_sucks', 'lights_dim', 'los_angeles', 'maintained_grocery', 'manager_enforce', 'masks_refuse', 'meet_classes', 'memberships_technically', 'minded_giving', 'morning_favorite', 'monthly_provide', 'nation_beginning', 'needing_cleaned', 'negatives_obvious', 'office_playing', 'online_daunting', 'overnight_clears', 'piling_fees', 'policy_signs', 'pools_organized', 'potential_skin', 'preening_posing', 'premature_aging', 'professional_fat', 'provide_bank', 'rates_equipments', 'reasonable_operating', 'rock_attentive', 'rusted_sprayed', 'saturday_morning', 'saturdays_sundays', 'shell_larger', 'simply_amaze', 'skin_damage', 'sleep_strip', 'spa_works', 'string_tops', 'stupid_pictures', 'summer_quality', 'switch_automated', 'task_cancelled', 'theyve_stand', 'thoughtfully_answers', 'tired_string', 'tops_clotgre', 'towns_understaffed', 'travelling_rates', 'unsure_mens', 'walking_miles', 'wall_immediately', 'wearing_masks', 'wondering_patrons', 'worked_coming', 'affordable_extremely', 'angeles_san', 'asked_minded', 'attach_rack', 'barbell_attach', 'barbell_bench', 'barbells_lead', 'bathrooms_millions', 'bath_rooms_needing', 'bed_ten', 'benches_barbell', 'body_builders', 'body_enhancement', 'boutique_style', 'cancelled_easiest', 'cash_credit', 'catches_kidding', 'chair_couple', 'closed_thursday', 'clothes_traveling', 'commitments_catc hes', 'contract_cancelling', 'country_boyfriend', 'crowded_feeling', 'crowded_wondering', 'customers_services', 'cu_stomers_wearing', 'days_notice', 'easiest_transaction', 'employee_sight', 'excellent_condition', 'experts_knowledgeable', 'extremely_annoying', 'freaked_boyfriend', 'guys_barely', 'guys_muscles', 'including_manager', 'including_smith']

# Generate bigrams from the original tokens
token_bigrams = list(ngrams(all_tokens, 2))

# Filter out bigrams that do not exist in the original tokenized text
valid_bigrams = []
for bigram in bigram_vocab:
    word1, word2 = bigram.split('_')

    # Check if the bigram is in the list of generated bigrams
    if (word1, word2) in token_bigrams:
        valid_bigrams.append(bigram)

# Update the final_gmap_dict with only valid bigrams
final_gmap_dict[gmap_id] = unigram_vocab + valid_bigrams

print(valid_bigrams)
print(len(valid_bigrams))

_monopolizing', 'grunting_rule', 'hassles_judgement', 'head_rusted', 'healthier_active', 'heart_attacks', 'ignore_rqquest', 'immediately_freaked', 'include_privacy', 'info_letting', 'intervention_acknowledging', 'jump_hoops', 'letting_piling', 'lifestyle_grunting', 'lighting_sucks', 'lights_dim', 'los_angeles', 'maintained_grocery', 'manager_enforce', 'masks_refuse', 'meet_classes', 'memberships_technically', 'minded_giving', 'morning_favorite', 'monthly_pr ovide', 'nation_beginning', 'needing_cleaned', 'negatives_obvious', 'office_playing', 'online_daunting', 'overnight_clears', 'piling_fees', 'policy_signs', 'pools_organized', 'potential_skin', 'preening_posing', 'premature_aging', 'professional_fat', 'provide_bank', 'rates_equipments', 'reasonable_operating', 'rock_attentive', 'rusted_sprayed', 'saturday_morning', 'saturdays_sundays', 'shell_larger', 'simply_amaze', 'skin_damage', 'sleep_strip', 'spa_works', 'string_tops', 'stupid_pictures', 'summer_quality', 'switch_automated', 'task_cancelled', 'theyve_stand', 'thoughtfully_answers', 'tired_string', 'tops_clotgre', 'towns_understaffed', 'travelling_rates', 'unsure_mens', 'walking_miles', 'wall_immediately', 'wearing_masks', 'wondering_patrons', 'worked_coming', 'affordable_extremely', 'angeles_s an', 'asked_minded', 'attach_rack', 'barbell_attach', 'barbell_bench', 'barbells_lead', 'bathrooms_millions', 'bath_rooms_needing', 'bed_ten', 'benches_barbell', 'body_builders', 'body_enhancement', 'boutique_style', 'cancelled_easiest', 'cash_credit', 'catches_kidding', 'chair_couple', 'closed_thursday', 'clothes_traveling', 'commitments_catc hes', 'contract_cancelling', 'country_boyfriend', 'crowded_feeling', 'crowded_wondering', 'customers_services', 'cu_stomers_wearing', 'days_notice', 'easiest_transaction', 'employee_sight', 'excellent_condition', 'experts_knowledgeable', 'extremely_annoying', 'freaked_boyfriend', 'guys_barely', 'guys_muscles', 'including_manager', 'including_sm ith']

142
```

## Checkpoint 3

Date**Time**: 2024-08-25 18:12

**Challenge:** When checking if the 200 bigram is in the text, the challenge is that which text should we use, the original text or the one after remove stopwords? The output gives 50 words out of 200 if we use original text, it gives 142 if we use the one after remove stopwords. From the quality of output, the 50 words seems more meaningful and we can abstract more information by only look at the bigrams. For the 142, some of them make sense and some of them is not that clear. Zihan think that we need to use the one after stemming and cleaning, I think we need to go with original text. Thus, we had some discussion on that. Now, we decide to use the one after remove stopwords because it not only includes the same 50 from original text, but also gives more words to analysis. Sometime more can be better to analysis.

**Contribution:** Ruiwen Chen

**Proof:**

```
# Iterate through tokenizer_dict to get original text
for gmap_id, da in tokenizer_dict.items():
    original_text_tokens = []

    # Gather all tokens from the tokenizer_dict for this gmap_id
    for date, tokens in da.items():
        original_text_tokens.extend(tokens)

    # Generate bigrams from the original tokens
    token_bigrams = list(ngrams(original_text_tokens, 2))

    # Filter out bigrams that do not exist in the original tokenized text
    valid_bigrams = []
    for bigram in bigram_vocab:
        word1, word2 = bigram.split('_')

        # Check if the bigram is in the list of generated bigrams
        if (word1, word2) in token_bigrams:
            valid_bigrams.append(bigram)

    # Update the final_gmap_dict with only valid bigrams
    #final_gmap_dict[gmap_id] = unigram_vocab + valid_bigrams
#print(original_text_tokens)
#print(valid_bigrams)
print(len(valid_bigrams))

['active_lifestyle', 'answers_questions', 'automated_system', 'bank_account', 'beginning_exerciser', 'broom_closet',
 'changing_stalls', 'classes_pools', 'code_tired', 'common_courtesy', 'daunting_task', 'dress_code', 'early_saturday',
 'effective_comparing', 'extended_periods', 'groups_monopolizing', 'grunting_rule', 'healthier_active', 'heart_attacks',
 'immediately_freaked', 'include_privacy', 'lighting_sucks', 'los_angeles', 'overnight_clears', 'pools_organized',
 'premature_aging', 'reasonable_operating', 'saturday_morning', 'simply_amaze', 'skin_damage', 'spa_works',
 'string_tops', 'summer_quality', 'thoughtfully_answers', 'barbell_bench', 'bathrooms_millions', 'bathrooms_needing',
 'body_builders', 'body_enhancement', 'boutique_style', 'closed_thursday', 'commitments_catches', 'crowded_feeling',
 'crowded_wondering', 'customers_services', 'days_notice', 'easiest_transaction', 'excellent_condition', 'extremely_irritating',
 'including_manager']
50
```

```

# Generate bigrams from the original tokens
token_bigrams = list(ngrams(all_tokens, 2))

# Filter out bigrams that do not exist in the original tokenized text
valid_bigrams = []
for bigram in bigram_vocab:
    word1, word2 = bigram.split('_')

    # Check if the bigram is in the list of generated bigrams
    if (word1, word2) in token_bigrams:
        valid_bigrams.append(bigram)

# Update the final_gmap_dict with only valid bigrams
#final_gmap_dict[gmap_id] = unigram_vocab + valid_bigrams

print(valid_bigrams)
print(len(valid_bigrams))

_monopolizing', 'grunting_rule', 'hassles_judgement', 'head_rusted', 'healthier_active', 'heart_attacks', 'ignore_rq
euest', 'immediately_freaked', 'include_privacy', 'info_letting', 'intervention_acknowledging', 'jump_hoops', 'lett
ing_piling', 'lifestyle_grunting', 'lighting_sucks', 'lights_dim', 'los_angeles', 'maintained_grocery', 'manager_e
nforce', 'masks_refuse', 'meet_classes', 'memberships_technically', 'minded_giving', 'morning_favorite', 'mothly_pr
ovide', 'nation_beginning', 'needing_cleaned', 'negatives_obvious', 'office_playing', 'online_daunting', 'overnight
_clears', 'piling_fees', 'policy_signs', 'pools_organized', 'potential_skin', 'preening_posing', 'premature_aging',
'professional_fat', 'provide_bank', 'rates_equipments', 'reasonable_operating', 'rock_attentive', 'rusted_sprayed',
'saturday_morning', 'saturdays_sundays', 'shell_larger', 'simply_amaze', 'skin_damage', 'sleep_strip', 'spa_works',
'string_tops', 'stupid_pictures', 'summer_quality', 'switch_automated', 'task_cancelled', 'theyve_stand', 'thoughtf
ully_answers', 'tired_string', 'tops_clotgre', 'towns_understaffed', 'travelling_rates', 'unsure_mens', 'walking_mi
les', 'wall_immediatly', 'wearing_masks', 'wondering_patrons', 'worked_coming', 'affordable_extremely', 'angeles_s
an', 'asked_minded', 'attach_rack', 'barbell_attach', 'barbell_bench', 'barbells_lead', 'bathrooms_millions', 'bath
rooms_needing', 'bed_ten', 'benches_barbell', 'body_builders', 'body_enhancement', 'boutique_style', 'cancelled_eas
iest', 'cash_credit', 'catches_kidding', 'chair_couple', 'closed_thursday', 'clothes_traveling', 'commitments_catc
hes', 'contract_cancelling', 'country_boyfriend', 'crowded_feeling', 'crowded_wondering', 'customers_services', 'cu
stomers_wearing', 'days_notice', 'easiest_transaction', 'employee_sight', 'excellent_condition', 'experts_knowledge
able', 'extremely_irritating', 'freaked_boyfriend', 'guys_barely', 'guys_muscles', 'including_manager', 'including_sm
ith']

142

```

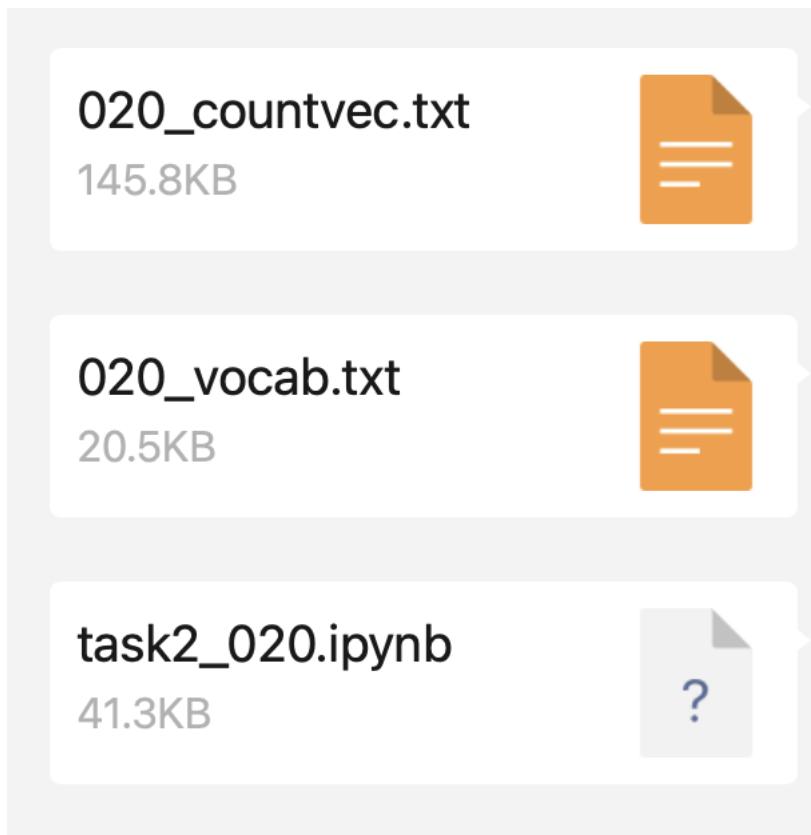
## Checkpoint 4

**Date/Time:** 2024-08-27 18:12

**Content:** Finish all the codes of task 2.

**Task 2 State:** task2\_020.ipynb can output the required .txt files, which are able to pass task2\_test.py. task2\_020.ipynb notebook has been handed over to Zihan Yin for some additional adjustments.

**Contribution:** Ruiwen Chen



**Proof:**

## Checkpoint 5

**DateTime:** 2024-08-29 6:41

**Content:** Make minor changes to task2\_020.ipynb.

Consider that the output of task 2 is wrong.

Find out some issues existing in task 2

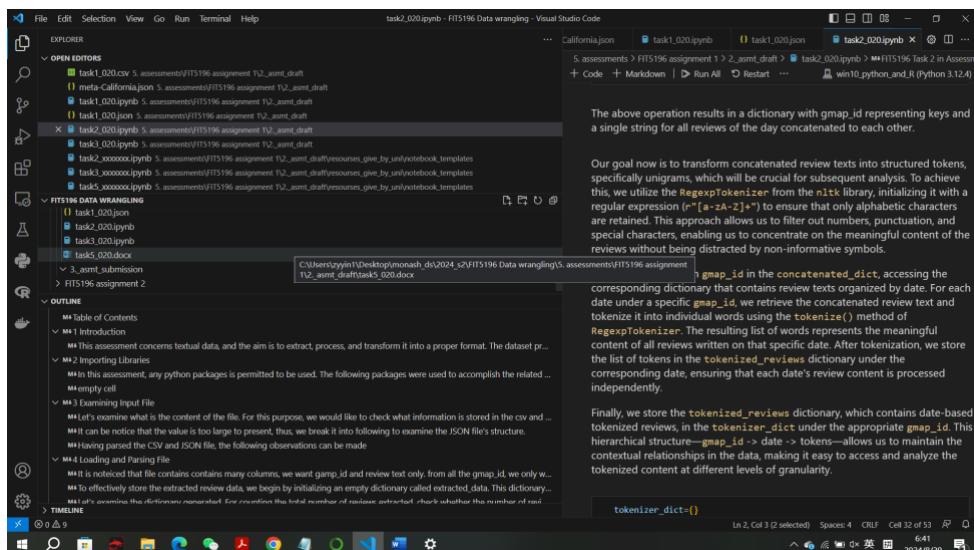
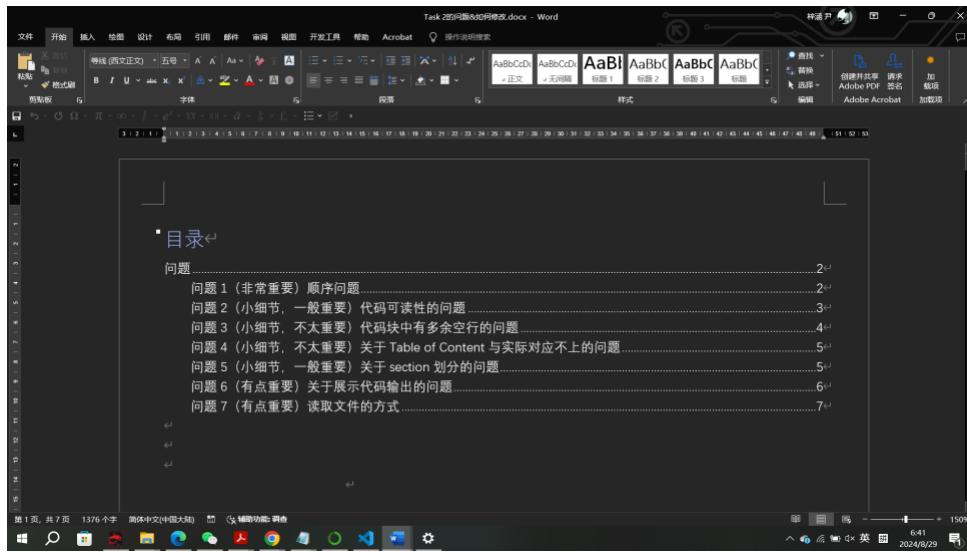
Create a word document to explain the issues and details that needed to be modified

**State:** Hand over the word document explaining the issues together with the task2\_020.ipynb file to Ruiwen Chen.

**Contribution:** Zihan Yin

**Proof:**

<Here's a screenshot of the word document explaining the issues. Please ignore the language used in this document to describe the issues>



## Checkpoint 6

**Date**: 2024-08-29 7:00

**Content**: Frequency list are all 1.

**Challenge**: The frequency in `countvec.txt` are all 1, which is meaningless for analysis and there must be something wrong. I checked the code again and identified that I used `set()` when create '`all_tokens`' and '`unigram_list`'. `Set()` function will filter out the repetition automatically, result in no repetition. The solution I apply to this is to change the `set()` to lists `[]`.

**Contribution**: Ruiwen Chen

## Proof:

```
0x54cb977215fdb773:0xa8189ecf8ad14096, 105:1, 223:1, 645:1, 722:1, 962:1, 1000:1, 1051:1,  
1144:1, 1180:1, 1701:1, 1792:1, 2178:1, 2196:1, 2377:1, 2382:1, 2567:1, 2628:1, 2663:1, 2716:1,  
2744:1, 2891:1, 2968:1, 3274:1, 3315:1, 3518:1, 3656:1, 3666:1, 3692:1, 3818:1, 3926:1, 4341:1,  
4351:1, 4429:1, 4562:1, 4568:1, 4619:1, 4652:1, 4701:1, 4749:1, 4782:1, 4790:1, 4886:1, 4908:1,  
4938:1, 5017:1, 5020:1, 5223:1, 5331:1, 5364:1, 5514:1, 5535:1, 5585:1, 5630:1, 5751:1, 5860:1,  
6001:1, 6108:1, 6148:1, 6171:1, 6445:1, 6501:1, 6537:1, 6632:1, 6673:1, 6705:1, 6725:1, 6758:1,  
6829:1, 6867:1, 6969:1, 7009:1, 7032:1, 7144:1, 7458:1, 7490:1, 7615:1, 7638:1, 7651:1, 7668:1,  
7671:1, 7869:1, 7875:1, 7973:1, 8102:1, 8112:1, 8280:1, 8325:1, 8346:1, 8362:1, 8424:1, 8459:1,  
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9418:1, 9596:1, 9720:1, 9801:1, 9819:1, 9826:1, 9896:1, 9997:1, 10104:1, 10172:1, 10299:1, 10305:1,  
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## After adjustment, frequency are not all 1.

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# Task 3

## Checkpoint 1

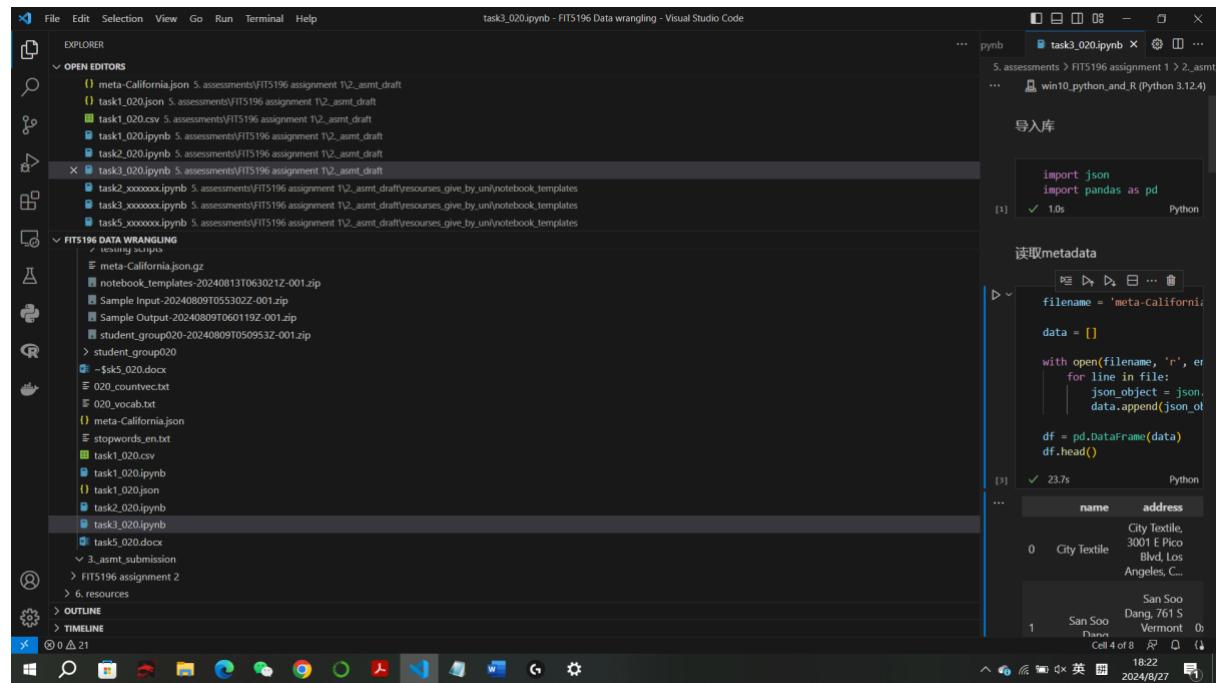
**DateTime:** 2024-08-27 18:12

**Content:** None

**State:** task3\_020.ipynb is just started, and loading the metadata.

**Contribution:** Zihan Yin

### Proof:



```
task3_020.ipynb - FIT5196 Data wrangling - Visual Studio Code
```

```
import json
import pandas as pd
```

```
filename = 'meta-California.json'
data = []
with open(filename, 'r') as f:
    for line in f:
        json_object = json.loads(line)
        data.append(json_object)

df = pd.DataFrame(data)
df.head()
```

	name	address
0	City Textile	3001 E Pico Blvd, Los Angeles, C...
1	San Soo Dang	761 S Vermont Dr., D...

## Checkpoint 2

**DateTime:** 2024-08-29 18:45

**Content:** None

### Task 3 State:

The Step 1 Understanding the sample google review data is halfway done. Step 2 and step 3 have already conceived some insights.

**Contribution:** Zihan Yin

**Proof:**

## Checkpoint 3

**Date**: 2024-08-30 15:34

**Content**: None

**Task 3 State**:

Step 1 & 2 has been finished, and step 3 is half-way done.

**Contribution**: Zihan Yin

**Proof:**

# Checkpoint 4

**DateTime:** 2024-08-30 18:25

**Content:** None

**Task 3 State:**

All done.

**Contribution:** Zihan Yin

The screenshot shows a Visual Studio Code interface with multiple tabs open. The active tab is titled "task3\_02.ipynb - FIT5196 Data wrangling - Visual Studio Code". The code editor contains Python code for data wrangling, specifically calculating the number of times each business was recommended based on the 'relative\_results' column and storing this as the 'times\_recommended' column. The code uses pandas operations like `apply`, `fillna`, and `drop` to manipulate the data. Below the code, a Jupyter Notebook cell displays a DataFrame named 'metadata'. The DataFrame has columns: gmap\_id, name, avg\_rating, num\_of\_reviews, MISC, location, city, primary\_category, price\_level, and ope. Two rows of data are shown:

gmap_id	name	avg_rating	num_of_reviews	MISC	location	city	primary_category	price_level	ope
0x0c2c58c0e3cf16dd529eda72874fd9	City Textile	4.5	6	None	[34.0188913, -118.2152898]	Los Angeles	Textile exporter	Unknown	Price Level
0x0c2c77e3b73d330xbdb58662xa97d49	San Soo Dang	4.4	18	{Service: options: Takeout, Dine in, Delivery}	[34.0580917, -118.2921295]	Los Angeles	Korean restaurant	Unknown	Price Level

Proof: