▼ Nodes

Node Label	Properties	Count
Business: Node to represent a single business.	business_id: Business unique identifier. business_name: Business name. business_type: Business type (grocery_store, fast_food, farmers_market, bakery). location: Business lat/lon. address: Business address. rating: avg. customer rating. price_level: rating for avg cost of goods.	
BlockGroup: Node that represents a census block group.	blockgroup_id: Block group id. Equal to the ctblockgroup in SANDAG. census_tract: Census tract the block group belongs to. block_group: Block group of census tract. object_id: Unique identifier for the shape. geometry: WKB polygon shape	2057
Zipcode: Node that represents a zip code.	zip: The 5-digit number identifying a zip code. geometry: WKB polygon shape	
City: Node that represents a city (provided by "Administrative Topology").	city_id: City unique identifier. city_name: City name. state_name: State city belongs to. county: County city belongs to. is_unincorporated: Boolean indicating the incorporation status of the city(or place).	
Neighborhood : Node that represents a neighborhood (provided by "Administrative Topology").	neighborhood_id : Unique identifier for each neighborhood. neighborhood_name : Name of the neighborhood.	
TotalPopulation : Node containing the total population level of the block group.	level: 2024 Total population level.	
PopulationGrowth: Node to categorize each block group based on its growth rate.	growth_rate : 2024-2029 Population: Compound Annual Growth Rate category	
AgeGroup: Node containing age group population representation of block groups.	group : An age group range, representation : The age groups population representation.	
EducationLevel : Node containing education level population representation of block groups.	level: The education level. representation: The education level population representation.	
WealthIndex: Node categorizing the wealth index of block groups.	category : The total wealth index category.	
CrimeIndex: Node categorizing crime index of block groups.	category : The total crime index category.	
FastFoodSpendingIndex: Node categorizing spending levels at fast food places, including take-out and delivery.	category : The fast food spending index category.	

▼ Edges

Relationship Type	Properties	Source Node	Target Node
HAS_NEIGHBOR	neighbor_type: "City" "Neighborhood".	City Neighborhood	→ City → Neighborhood
HAS_NEARBY	nearby_type: "City" "Neighborhood".	City Neighborhood Neighborhood	<pre>→ City → Neighborhood → City</pre>
LOCATED_IN		Business Business	→ BlockGroup → Zipcode
HAS_NEIGHBORHOOD		City	→ Neighborhood
IS_WITHIN	<pre>containment_type : "Full" "Partial"</pre>	City Neighborhood BlockGroup	→ Zipcode→ Zipcode→ Zipcode
HAS_ENRICHMENT		BlockGroup	→ TotalPopulation → PopulationGrowth → AgeGroup → EducationLevel → WealthIndex → CrimeIndex → FastFoodSpendingIndex

Geo-enrichment Mapping

Enrichments +

Aa Name			≡ Use	■ Nodes		■ Values
TOTPOP_CY	2024 Total Population	Demographic	categorize each block group based on its population	TotalPopulation	level	"low": fewer than 1,000 residents "Medium: between 1,000 and 2,000 residents "High": more than 2,000 residents
POPGRWCYFY	2024-2029 Population: Compound Annual Growth Rate	Demographic	categorize each block group based on its growth rate	PopulationGrowth	growth_rate	"Negative": less than 0% (negative rate) "Low": 0% to 1% annually "Moderate": 1% to 2% annually "High": 2% to 3% annually "Very High": greater than 3% annually
male0 - male 85, OPEN fem85	2024 Male Population Age (© (§) & 2024 Female Population Age 0-85	Demographic	compute average age for each block group, categorize each block group based on its average age	AgeGroup	group representation	group: "0-4", "5-14", "15-24", "25-44", "45-64", "66+" representation: "Very Low": 0% to 5%. "Low": 5% to 10%. "Moderate": 10% to 20%. "High": 20% to 30%. "Dominant": Over 30%.
withindxcy	2023 Wealth Index	Socioecono	categorize each block group into richest, upper middle, mid-class, lower middle, and poverty Normalized it from 0 to 1	WealthIndex	category	"Low": Index score between 0.0 and 0.2 "Lower-Middle": Index score between 0.2 and 0.4 Middle: Index score between 0.4 and 0.6 Upper-Middle: Index score between 0.6 and 0.8 "High": Index score between 0.8 and 1.0
NOHS_CY, SOMEHS_CY, HSGRAD_CY, GED_CY, SMCOLL_CY, ASSCDEG_CY, BACHDEG_CY, GRADDEG_CY, educbasecy	2024 Population Age 25+: Less than 9th Grade, 9-12th Grade/No Diploma , High School Diploma, GED/Alternative Credential, Some College/No Degree, Associate's Degree, Bachelor's Degree, Graduate/Professional Degree	Socioecono	compute the percentage of each block group in terms of Basic Education, Secondary Education, and Higher Education. basic_education_pct: Includes NOHS_CY, SOMEHS_CY, secondary_education_pct: Includes HSGRAD_CY, GED_CY, SMCOLL_CY, Higher_education_pct: Includes ASSCDEG_CY, BACHDEG_CY, GRADDEG_CY	EducationLevel	level representation	level: "Basic": Less than 9th Grade, 9-12th Grade/No Diploma. "Secondary": High School Diploma, GED/Alternative Credential, Some College/No Degree. "Higher": Associate's Degree, Bachelor's Degree, Graduate/Professional Degree. representation: "Very Low": 0% to 5%. "Low": 5% to 15%. "Moderate": 15% to 30%. "High": 30% to 50%. "Very High": Over 50%.
CRMCYTOTC OPEN	2024 Total Crime Index	Socioecono	categorize each block group from safest to most unsafe based on the Total Crime Index	CrimeIndex	category	"Safest": Index < 80 "Safe": Index 80-119 "Moderate": Index 120-199 "Unsafe": Index 200-499 "Most Unsafe": Index ≥ 500
x1133a, x1138a, x1148a	spending (\$) on lunch, dinner, breakfast at fast food/take- out/deliv	Spending	categorize spending levels at fast food places, including take-out and delivery combine x1133a, x1138a, x1148a first, then Normalized it from 0 to 1	FastFoodSpendingIndex	category	"Occasional": Index score between 0.0 and 0.2 "Light spender": Index score between 0.2 and 0.4 "Regular": Index score between 0.4 and 0.6 "Enthusiast": Index score between 0.6 and 0.8 "Super Fan": Index score between 0.8 and 1.0

Data Transformations



Documentation for all data transformations required for constructing the knowledge graph (excluding any text processing and NLP operations).

▼ Pseudocode

1. Initialize Neo4j Environment

```
CONNECT to Graph
```

- 2. Extract and Transform Data from Sources
 - a. Block Groups

b. Neighborhoods & Cities

```
CONNECT to Postgres server
LOAD city_neighborhoods table
LOAD community_neighborhoods table
CREATE_NODES:
    # City
    FOR EACH UNIQUE city DO
        CREATE_NODE(City(city_id, city_name, state_name, county))
    END FOR
    # Neighborhood
    FOR EACH UNIQUE neighborhood DO
        CREATE_NODE(Neighborhood(neighborhood_id, neighborhood_name))
    END FOR
CREATE_RELATIONSHIPS:
    FOR EACH city in graph DO
        neighboring_cities = GET_NEIGHBORING_CITIES(city)
        nearby_cities = GET_NEARBY_CITIES(city)
        # City -> City (HAS_NEIGHBOR)
        FOR EACH neighbor_city in neighboring_cities DO
```

```
CREATE_EDGE(HAS_NEIGHBOR, city, neighbor_city)
   END FOR
   # City -> City (HAS_NEARBY)
   FOR EACH nearby_city in nearby_cities DO
        CREATE_EDGE(HAS_NEARBY, city, nearby_city)
    END FOR
END FOR
FOR EACH neighborhood in graph DO
    neighboring_neighborhoods = GET_NEIGHBORING_NEIGHBORHOODS(neighborhood)
    nearby_neighborhood = GET_NEARBY_NEIGHBORHOODS(neighborhood)
   nearby_cities = GET_NEARBY_CITIES(neighborhood)
   # Neighborhood -> Neighborhood (HAS_NEIGHBORHOOD)
   FOR EACH neighbor_neighborhood in neighboring_neighborhoods DO
        CREATE_EDGE(HAS_NEIGHBORHOOD, neighborhood, neighbor_neighborhood)
   END FOR
   # Neighborhood -> Neighborhood (HAS_NEARBY)
   FOR EACH nearby_neighborhood in nearby_neighborhoods DO
        CREATE_EDGE(HAS_NEARBY, neighborhood, nearby_neighborhood)
   END FOR
   # Neighborhood -> City (HAS_NEARBY)
   FOR EACH nearby_city in nearby_cities DO
        CREATE_EDGE(HAS_NEARBY, neighborhood, nearby_city)
   END FOR
END FOR
```

c. Zipcodes

```
CONNECT to Postgres server
OPEN csv_file("zipcodes.csv") as CSV_FILE:
 FOR EACH row in CSV_FILE:
   zipcode = row['ZIP']
    geometry = row['the_geom'].wkb
       CREATE_NODE(ZipCode(zipcode_number, geometry))
    CREATE_RELATIONSHIPS:
       FOR EACH zipcode in GRAPH DO:
            # Decode WKB to geometry objects using Shapely library
          zipcode_polygon = wkb.loads(zipcode['geometry'])
          FOR EACH block_group in GRAPH DO:
              blockgroup_polygon = wkb.loads(block_group['geometry'])
              IF do_polygons_intersect(zipcode_polygon, blockgroup_polygon):
                  # Blockgroup -> ZipCode (IS_WITHIN)
                  CREATE_EDGE(IS_WITHIN, block_group, zipcode)
            zipcode_number = zipcode['zipcode_number']
            FOR EACH neighborhood in GRAPH DO:
                IF zipcode_number IN neighborhood['zipcodes']
                    # Neighborhood -> ZipCode (IS_WITHIN)
                    CREATE_EDGE(IS_WITHIN, neighborhood, zipcode)
```

d. Enrichments

```
# Connect to Postgres server and load the necessary table
CONNECT to Postgres server
# Load table containing GeoProfile data (assuming it's called geoprofiles_table)
LOAD bgs_sd_imp
# Create nodes for each attribute category in GeoProfile
FOR EACH ROW IN bgs_sd_imp RESULT DO
         # Total Population Node
         CREATE NODE(
                  type = "TotalPopulation",
                  total_population = ROW["TOTPOP_CY"]
         )
         # Population Growth Category Node
         CREATE_NODE(
                  type = "PopulationGrowthCategory",
                  population_growth_category = CATEGORIZE(ROW["POPGRWCYFY"], {
                           "Negative growth": ROW["POPGRWCYFY"] < 0,
                           "Low growth": 0 <= ROW["POPGRWCYFY"] < 1,
                           "Moderate growth": 1 <= ROW["POPGRWCYFY"] < 2,
                           "High growth": 2 <= ROW["POPGRWCYFY"] < 3,
                           "Very high growth": ROW["POPGRWCYFY"] >= 3
                 })
         )
         # Average Age Node
         CREATE_NODE(
                  type = "AverageAge",
                  average_age = CALCULATE_AVERAGE_AGE(ROW["male0"]...ROW["male85"], ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["fem0"]...ROW["f
85"])
        )
         # Wealth Index Node
         CREATE NODE(
                  type = "WealthIndex",
                 wealth_index = CATEGORIZE(ROW["wlthindxcy"], {
                           "poverty": 0.0 <= ROW["wlthindxcy"] < 0.2,
                           "lower-middle": 0.2 <= ROW["wlthindxcy"] < 0.4,
                           "mid-class": 0.4 <= ROW["wlthindxcy"] < 0.6,</pre>
                           "upper-middle": 0.6 <= ROW["wlthindxcy"] < 0.8,
                           "richest": 0.8 <= ROW["wlthindxcy"] <= 1.0
                 })
         )
         # Crime Index Category Node
         CREATE_NODE(
                  type = "CrimeIndexCategory",
                  crime_index_category = CATEGORIZE(ROW["CRMCYTOTC"], {
                           "Safest": ROW["CRMCYTOTC"] < 80,
                            "Safe": 80 <= ROW["CRMCYTOTC"] < 120,
                            "Moderate": 120 <= ROW["CRMCYTOTC"] < 200,
                            "Unsafe": 200 <= ROW["CRMCYTOTC"] < 500,
                           "Most unsafe": ROW["CRMCYTOTC"] >= 500
                 })
         )
```

```
# Spending Category Node
    CREATE_NODE(
        type = "SpendingCategory",
        spending_category = CATEGORIZE(NORMALIZE(ROW["x1133a"] + ROW["x1138a"] + ROW["x1148a"]),
{
            "Occasional": 0.0 <= value < 0.2,
            "Light spender": 0.2 <= value < 0.4,
            "Regular": 0.4 <= value < 0.6,
            "Enthusiast": 0.6 <= value < 0.8,
            "Super Fan": 0.8 <= value <= 1.0
       })
    )
    # Education Level Node
    CREATE NODE(
       type = "EducationLevel",
       basic_education_pct = CALCULATE_PERCENTAGE(ROW["NOHS_CY"], ROW["SOMEHS_CY"]),
       secondary_education_pct = CALCULATE_PERCENTAGE(ROW["HSGRAD_CY"], ROW["GED_CY"], ROW["SMC
OLL_CY"]),
        higher_education_pct = CALCULATE_PERCENTAGE(ROW["ASSCDEG_CY"], ROW["BACHDEG_CY"], ROW["G
RADDEG_CY"])
   )
END FOR
# Establish relationships between BlockGroup and each separate attribute node
FOR EACH blockgroup IN sandag_layer_census_block_groups DO
   FIND BlockGroup node based on blockgroup_id
    # For each attribute node type, establish relationships if both nodes exist
    FOR EACH attribute_node IN [TotalPopulation, PopulationGrowthCategory, AverageAge, WealthInd
ex, CrimeIndexCategory, SpendingCategory, EducationLevel] DO
       FIND attribute_node based on matching criteria
       IF BlockGroup node exists AND attribute_node exists THEN
           CREATE RELATIONSHIP from BlockGroup to attribute node
            SET relationship attribute value to blockgroup's specific value (if applicable)
       END IF
   FND FOR
FND FOR
```

e. Businesses

```
CREATE_EDGE(LOCATED_IN, Business(business_id=business["business_id"]), BlockGroup)

END IF

END FOR

END FOR
```