Manhattan Airbnb Property Analysis

Project Overview

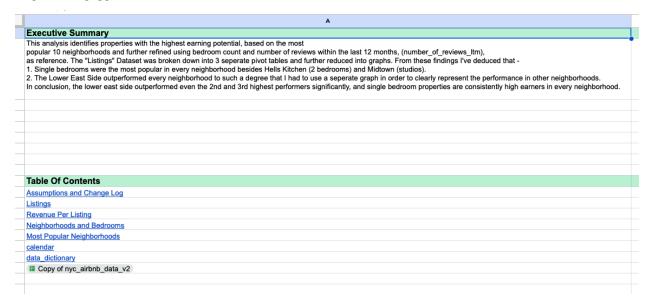
You've been hired to help a client analyze the Manhattan vacation rental market. The client wants guidance on which property types to invest in, and your task is to analyze Airbnb data for actionable insights.

This analysis answers two key questions:

- 1. Which neighborhoods and property sizes (number of bedrooms) are most attractive for vacation rentals?
- 2. How much money did these listings generate?

The project leverages real NYC Airbnb data, focusing on data cleaning, pivot table analysis, and revenue estimation to provide a data-driven roadmap for maximizing returns in the vacation rental sector.

Raw Data



Data Cleaning

Before diving into the analysis, the data was carefully cleaned to ensure accuracy and consistency:

- The neighborhood column was standardized for consistent capitalization and removal
 of trailing spaces, with cleaned values stored in a new column called
 neighborhood_clean.
- The bedrooms column was cleaned so that empty cells (representing studios) were set to zero, and the results were stored in a new column called bedrooms_clean.
- All cleaning steps were documented in a separate change log sheet, and a copy of the raw data was preserved for reference.

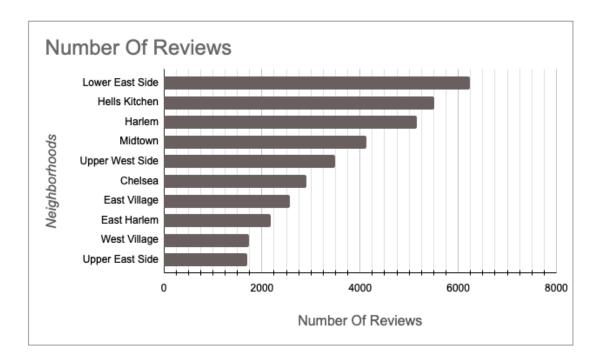
Analysis

1. Most Attractive Neighborhoods

To determine which neighborhoods are most attractive for vacation rentals, a pivot table was created using the number_of_reviews_ltm column as a proxy for rental frequency (since reviews are only left after a stay).

Top 10 Most Attractive Neighborhoods (by reviews):

Lower East Side | Hell's Kitchen | Harlem | (plus 7 more, see chart)



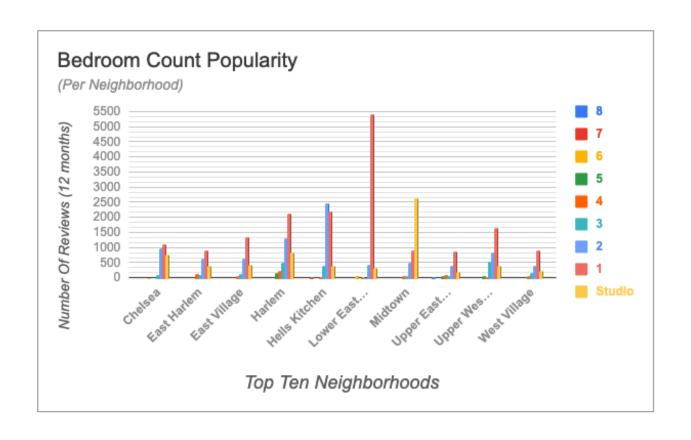
2. Most Popular Property Sizes

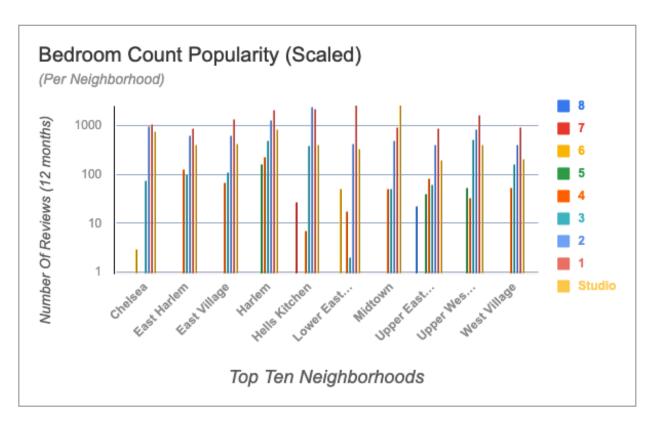
Next, the analysis focused on property size. A pivot table was used to determine which number of bedrooms were most popular among guests.

Top 3 Most Popular Property Sizes: Studios | 1-bedrooms | 2-bedrooms

These unit types consistently attracted the highest number of reviews, indicating strong guest preference, as seen below:

A	В	С	D	Е	F	G	Н	1	J∢	
Most Popular Neighborhoods Number Of Reviews (Last 12 Months)										
Bedroom Count	8	7	6	5	4	3	2	1	0	
Chelsea			3			76	958	1095	781	
East Harlem					131	101	639	899	405	
East Village					68	110	626	1332	436	
Harlem				165	228	494	1320	2126	824	
Hells Kitchen		27			7	394	2470	2204	404	
Lower East Side			51		18	2	429	5399	343	
Midtown					51	52	488	908	2629	
Upper East Side	23			41	85	61	407	885	194	
Upper West Side				55	33	521	847	1639	402	
West Village					53	162	402	907	211	
0										





3. Neighborhood-Specific Preferences

Manhattan neighborhoods show unique tastes in property sizes, impacting rental demand and income. Knowing these preferences helps investors make smarter choices.

Key Findings:

- 1. Harlem:
 - Most Popular: 1-bedrooms
 - Why? Affordable space for solo travelers and couples.
- 2. Lower East Side:
 - Most Popular: Studios and 1-bedrooms
 - Why? Great location for budget-conscious, short-term visitors.
- 3. Hell's Kitchen:
 - Most Popular: 1-bedrooms and 2-bedrooms
 - Why? Tourists and groups need more space.
- 4. Midtown:
 - Most Popular: Studios
 - Why? Cheaper alternative to hotels in a central spot.
- 5. Williamsburg (Brooklyn):
 - Most Popular: 1-bedrooms and 2-bedrooms
 - Why? Trendy area attracts longer stays and remote workers.

Investment Tips:

- Match property size to neighborhood. Invest in 1-bedrooms in Harlem or studios in Midtown.
- Avoid unpopular sizes. Don't oversupply 3+ bedrooms in Hell's Kitchen.
- **Consider a mix.** Some areas (like Lower East Side) do well with both studios and 1-bedrooms.

Insert chart here: "Most Popular Property Size by Neighborhood"

Next Steps:

- Adjust prices by neighborhood. Charge more for popular sizes in high-demand areas.
- Check out the competition. See how well different property sizes are doing in each area.

4. Revenue Analysis

To measure real-world earning potential, we analyzed revenue data from actual bookings. Each listing's income was calculated using the **calendar data**, tracking every night the property was rented. We created a new column called **revenue_earned**, which recorded the nightly **adjusted_price** if the property was booked, or **\$0** if it was vacant.

By summing up 30 days of bookings, we estimated **monthly revenue** for each listing, then projected **annual revenue** by multiplying by 12. This approach revealed a dramatic gap between the highest and lowest earners.

Key Findings:

- **Top 10 Listings:** The best-performing properties earned **\$29,940** in just **30 days**, projecting to **\$359,280** annually. These high earners typically had prime locations, strong reviews, and competitive pricing.
- Bottom 10 Listings: In contrast, the lowest earners struggled to generate meaningful income, with some barely covering costs. Poor photos, outdated listings, or inconvenient locations often held them back.

This stark difference highlights how **smart pricing**, **presentation**, **and location selection** can make or break an Airbnb investment. Investors should study top performers to replicate their success while avoiding the pitfalls of underperforming listings.





Results

- Lower East Side, Hell's Kitchen, and Harlem are the most attractive neighborhoods for Airbnb rentals.
- Studios, 1-bedrooms, and 2-bedrooms are the most popular property sizes.
- Neighborhood preferences vary, with 1-bedrooms being especially in demand in Harlem.
- There is a significant gap between the top and bottom earning listings, with the top listing generating over \$29,000 in a single month.

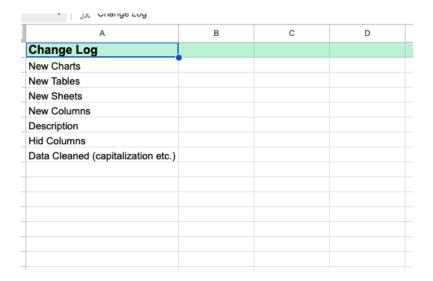
Conclusions

Manhattan's Airbnb market shows strong, data-driven trends in both neighborhood and property size preferences. Investors can maximize occupancy and revenue by focusing on the most in-demand areas and unit types. Revenue potential is significant, especially for well-located and appropriately sized properties, but performance varies widely between listings.

Recommendations

- Invest in the Lower East Side, Hell's Kitchen, or Harlem to capture high demand.
- **Prioritize studios and 1-bedroom units**, especially in neighborhoods where they are most popular.
- Regularly review and adjust pricing based on neighborhood trends and seasonality.
- Maintain high listing quality and monitor guest reviews to sustain strong rental performance.
- Analyze top and bottom performers to identify what drives success and avoid common pitfalls.

Change Log



Assumptions

Α	В	С	D	E	F	G	Н	I	J	K
Assumptions										
1	The dataset inc	ludes all active va	cation rental listin	ngs in the target n	narket for the ana	llysis period.				
2	Listings marked	as "active" are a	vailable for bookir	ng during the enti	re analysis period	i.				
3	Guest review co	ounts are used as	a proxy for booki	ngs, assuming re	view rates are co	nsistent across lis	stings.			
4	Revenue is cald	ulated as nightly	rate times booked	d nights; extra fee	es are excluded u	nless specified.				
5	Occupancy rate	is booked nights	divided by total a	vailable nights fo	r each listing.					
6	Listings are gro	uped by bedroom	count; listings wi	th missing or unc	lear bedroom dat	a are excluded fro	om property size a	analysis.		
7	Listings are ass	igned to neighbor	hoods based on a	address or coordi	inates; those with	missing or unclea	ar location data a	e excluded from	neighborhood ar	alysis.
8	Host activity is r	neasured by cale	ndar updates, res	ponse times, and	d recent listing cha	anges, if available).			
9	Nightly rates are	e treated as fixed								
10	Guest preference	ces are inferred fr								
11	The data period	is assumed to re								
12	Obvious data er	rors (e.g., imposs								
13	Revenue differe	nces are linked to	management pr	actices, listing qu	ality, and location	, based on availa	ble data.			
14	Listing visibility	is assumed to be	influenced by rev	iew count, respo	nse rate, and listi	ng updates.				
15	Local rental law									

Cleaned Raw Data

A ∢	▶ AC	♦ AK	▶ AM	AN	Top_NB		AP	AQ ◀	▶ BF	4
I	neighborhood_clean	bathrooms_text	bedrooms_clean	Top_Listing			Top_BD	revenue_earned	number_of_reviews_	
82638	East Village	1 bath	1		1	TRUE	TRUE	3549		;
9357	Hells Kitchen	1 bath	0		0	TRUE	FALSE	5250		(
57618	Hells Kitchen	1 bath	1		1	TRUE	TRUE	7500		
57754	Flatiron District	1 bath	1		0	FALSE	TRUE	9570		2
57874	East Harlem	1 bath	1		1	TRUE	TRUE	4320		5
15341	Lower East Side	1 bath	1		1	TRUE	TRUE	2729		
59014	Lower East Side	1 bath	1		1	TRUE	TRUE	5750		
59709	Chinatown	1 bath	2		0	FALSE	FALSE	8100		2
60164	Tribeca	1 bath	3		0	FALSE	FALSE	3500		
16974	East Harlem	2 baths	4		0	TRUE	FALSE	7350		2
60611	East Harlem	1 bath	1		1	TRUE	TRUE	3300		
107895	Upper West Side	1 bath	0		0	TRUE	FALSE	4980		
17037	East Village	1 bath	1		1	TRUE	TRUE	6600		
60680	East Village	1 bath	1		1	TRUE	TRUE	4800		
61509	Midtown	1 bath	1		0	TRUE	FALSE	3300		
62427	East Village	1 bath	1		1	TRUE	TRUE	3120		
19159	Harlem	1 bath	2		0	TRUE	FALSE	3190		
62891	East Village	1 bath	0		0	TRUE	FALSE	4725		
116551	Harlem	1 bath	0		0	TRUE	FALSE	2100		
63693	East Village	2 baths	2		0	TRUE	FALSE	11700		
23686	West Village	2 baths	3		0	TRUE	FALSE	14250		
123784	Harlem	1 bath	0		0	TRUE	FALSE	2739		3
126443	East Village	2 baths	2		0	TRUE	FALSE	12852		6
27644	Harlem	1 bath	1		1	TRUE	TRUE	4750		1
65425	Nolita	1 bath	2		0	FALSE	FALSE	3075		1
29683	Lower East Side	1 bath	1		1	TRUE	TRUE	5550		
66251	East Village	1 bath	0		0	TRUE	FALSE	9776		3
140973	East Village	Half-bath	2		0	TRUE	FALSE	6450		
67288	East Harlem	1 bath	1		1	TRUE	TRUE	3665		
144087	Harlem	1 bath	2		0	TRUE	FALSE	9584		
36703	West Village	1 bath	1		1	TRUE	TRUE	6000		
148825	Upper West Side	1 bath	2		0	TRUE	FALSE	5021		3
68974	Nolita	1 bath	3		0	FALSE	FALSE	17275		
69894	Upper West Side	1 bath	1		1	TRUE	TRUE	4800		
42300	Lower East Side	2 baths	2		0	TRUE	FALSE	19950		
71248	Chinatown	1 bath	1		0	FALSE	TRUE	4650		