

Data Loading and Preprocessing

First, we'll connect to the SQLite database and load the data into a pandas DataFrame.

	meeting_id	duration	attendees	salary
0	755788	0.75	1	250000.0
1	755788	0.75	1	163808.0
2	653454	0.25	1	250000.0
3	653454	0.25	1	135000.0
4	653454	0.25	1	163808.0

Calculate Total Cost and Total Cost per Meeting

Now, we'll add columns for total cost and total cost per meeting.

	meeting_id	duration	attendees	salary	total_cost	total_cost_per_meeting
0	755788	0.75	1	250000.0	187500.0	310356.0
1	755788	0.75	1	163808.0	122856.0	310356.0
2	653454	0.25	1	250000.0	62500.0	168452.0
3	653454	0.25	1	135000.0	33750.0	168452.0
4	653454	0.25	1	163808.0	40952.0	168452.0

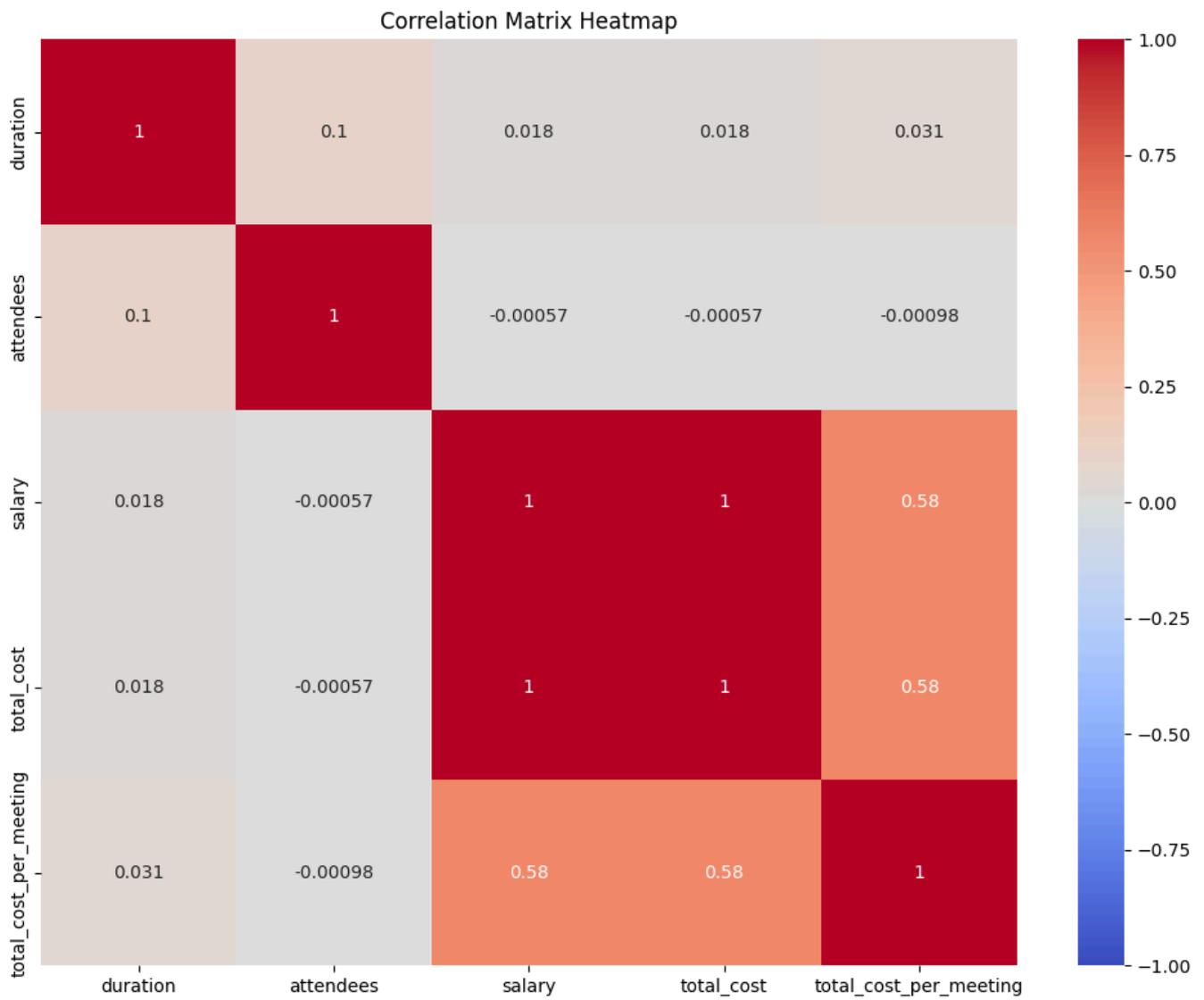
	meeting_id	duration	attendees	salary	total_cost	total_cost_per_meeting
0	755788	0.75	1	250000.0	187500.0	310356.0
1	755788	0.75	1	163808.0	122856.0	310356.0
2	653454	0.25	1	250000.0	62500.0	168452.0
3	653454	0.25	1	135000.0	33750.0	168452.0
4	653454	0.25	1	163808.0	40952.0	168452.0

Correlation Analysis

Let's calculate and visualize the correlation matrix for our numeric columns.

Correlation Matrix:

	duration	attendees	salary	total_cost	total_cost_per_meeting
duration	1.000000	0.100116	0.018141	0.018141	0.031437
attendees	0.100116	1.000000	-0.000566	-0.000566	-0.000981
salary	0.018141	-0.000566	1.000000	1.000000	0.577059
total_cost	0.018141	-0.000566	1.000000	1.000000	0.577059
total_cost_per_meeting	0.031437	-0.000981	0.577059	0.577059	1.000000



Outlier Detection and Removal

Now, let's identify and remove outliers using the Interquartile Range (IQR) method.

Q1 values:
duration 0.5
attendees 1.0
salary 60000.0
total_cost 60000.0
total_cost_per_meeting 195000.0
Name: 0.25, dtype: float64

Q3 values:
duration 1.0
attendees 6.0
salary 150000.0
total_cost 600000.0
total_cost_per_meeting 1060000.0
Name: 0.75, dtype: float64

IQR values:
duration 0.5
attendees 5.0
salary 90000.0
total_cost 540000.0
total_cost_per_meeting 865000.0
dtype: float64

Outliers detected:

0	False
1	False
2	False
3	False
4	False
...	
1801	False
1802	False
1803	False
1804	True
1805	False

Length: 1806, dtype: bool

DataFrame after removing outliers:

	meeting_id	duration	attendees	salary	total_cost	total_cost_per_meeting
0	755788	0.75	1	250000.0	187500.0	310356.0
1	755788	0.75	1	163808.0	122856.0	310356.0
2	653454	0.25	1	250000.0	62500.0	168452.0
3	653454	0.25	1	135000.0	33750.0	168452.0
4	653454	0.25	1	163808.0	40952.0	168452.0

Original dataset size: 1806

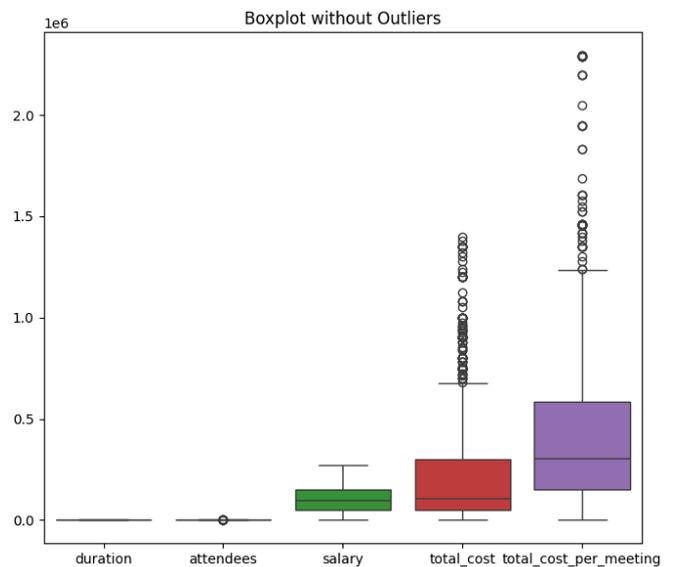
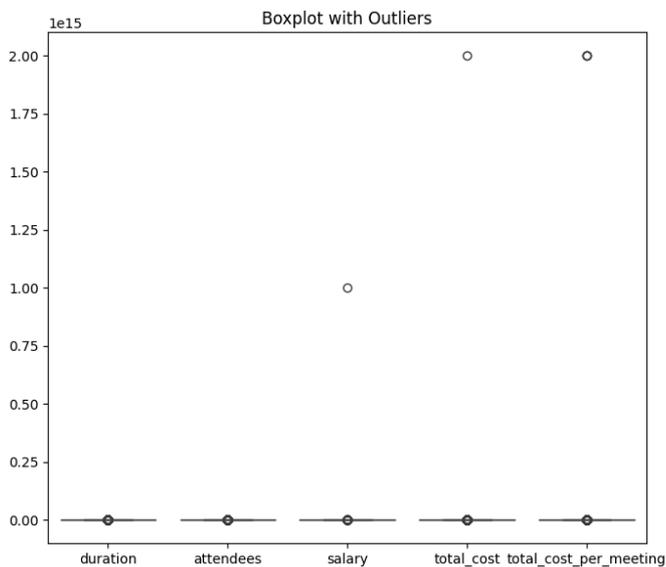
Dataset size after removing outliers: 1339

Min/Max values before outlier handling:

	duration	attendees	salary	total_cost	total_cost_per_meeting
min	0.25	0	0.000000e+00	0.000000e+00	0.000000e+00
max	8.00	1000000	1.000000e+15	2.000000e+15	2.000100e+15

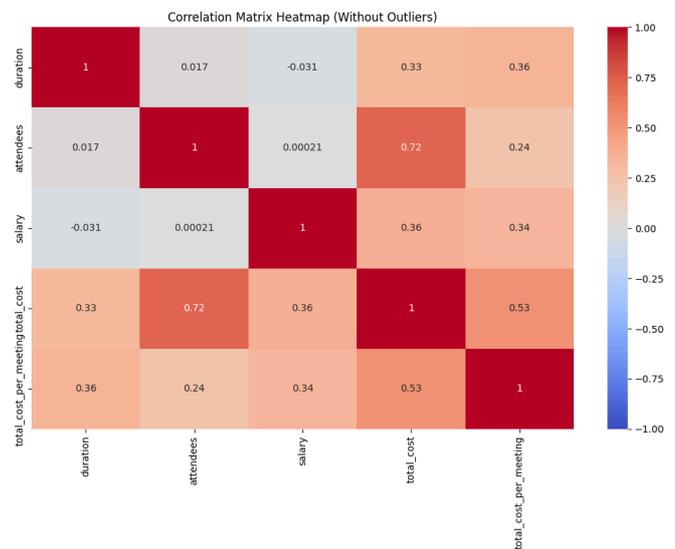
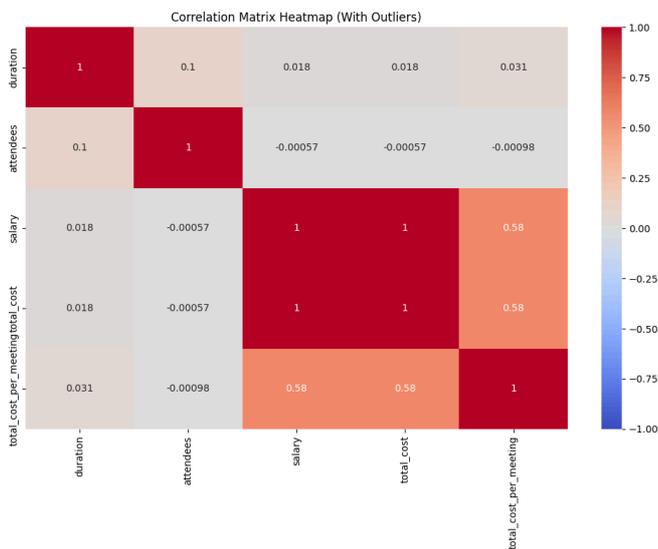
Min/Max values after outlier handling:

	duration	attendees	salary	total_cost	total_cost_per_meeting
min	0.25	0	0.0	0.0	0.0
max	1.75	13	270000.0	1400000.0	2293750.0



Correlation Analysis Without Outliers

Let's recalculate the correlation matrix after removing outliers and compare it with the original.



Correlation Matrix (With Outliers):

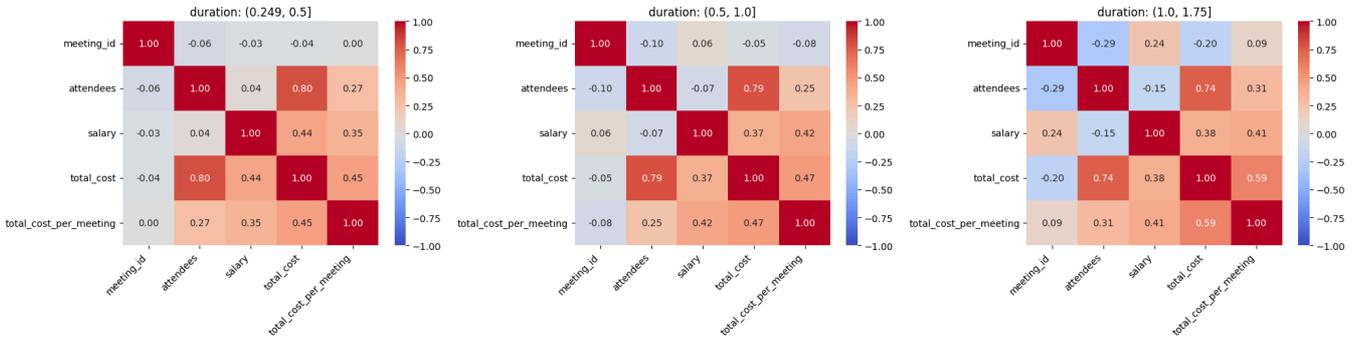
	duration	attendees	salary	total_cost	total_cost_per_meeting
duration	1.000000	0.100116	0.018141	0.018141	0.031437
attendees	0.100116	1.000000	-0.000566	-0.000566	-0.000981
salary	0.018141	-0.000566	1.000000	1.000000	0.577059
total_cost	0.018141	-0.000566	1.000000	1.000000	0.577059
total_cost_per_meeting	0.031437	-0.000981	0.577059	0.577059	1.000000

Correlation Matrix (Without Outliers):

	duration	attendees	salary	total_cost	total_cost_per_meeting
duration	1.000000	0.016571	-0.031110	0.326703	0.356491
attendees	0.016571	1.000000	0.000209	0.724696	0.242574
salary	-0.031110	0.000209	1.000000	0.361457	0.337871
total_cost	0.326703	0.724696	0.361457	1.000000	0.529158
total_cost_per_meeting	0.356491	0.242574	0.337871	0.529158	1.000000

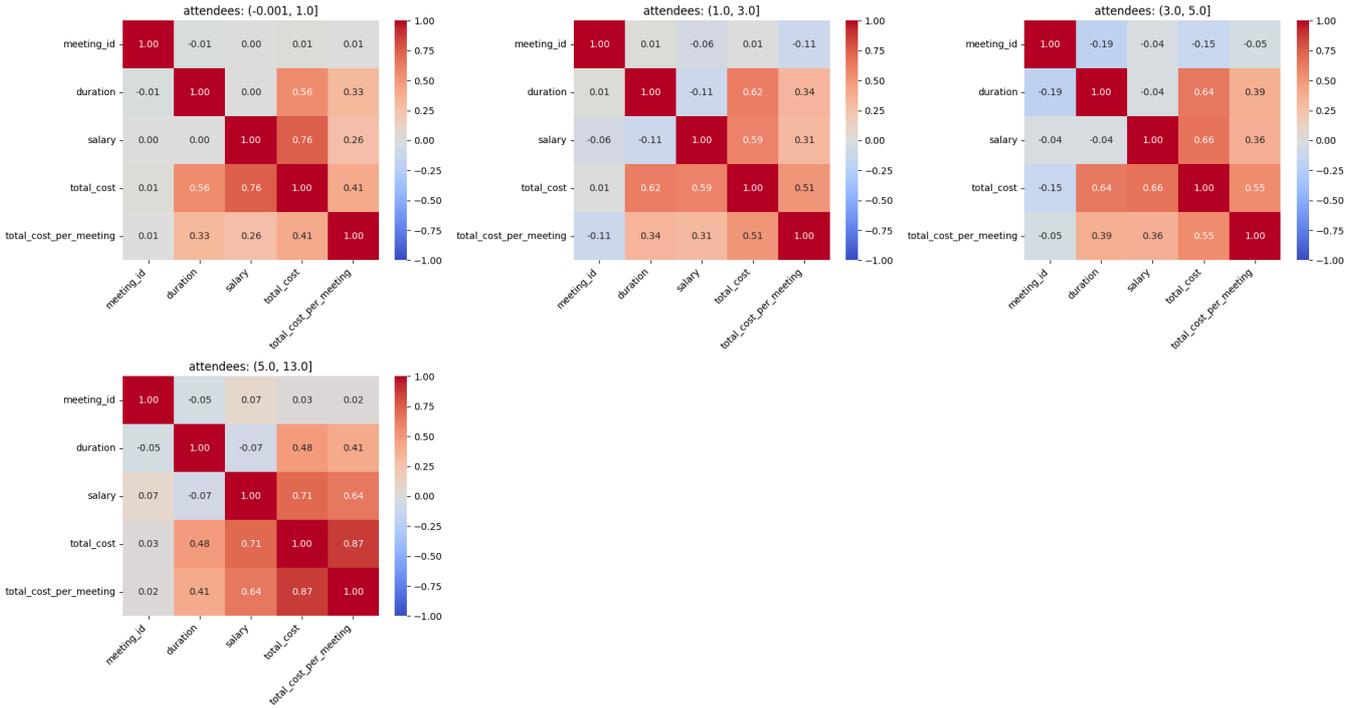
Analyzing correlations by duration

Correlation Heatmaps by duration



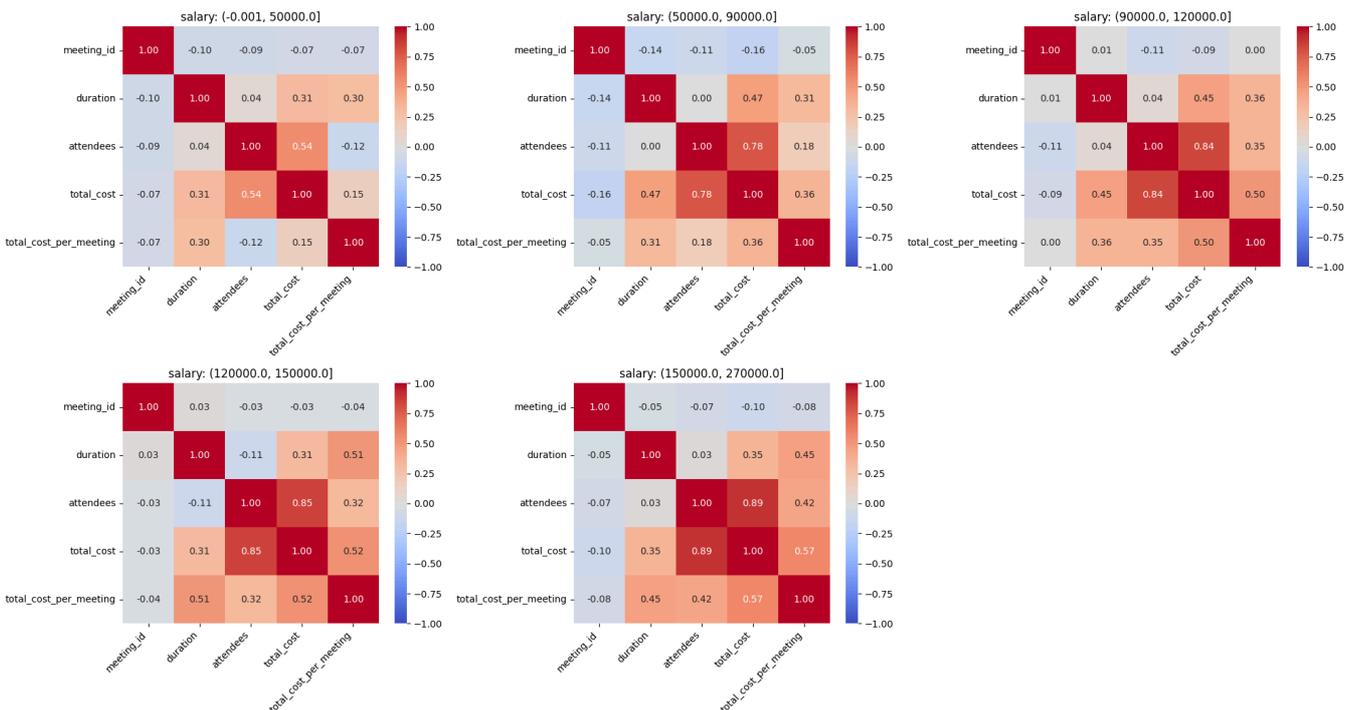
Analyzing correlations by attendees

Correlation Heatmaps by attendees



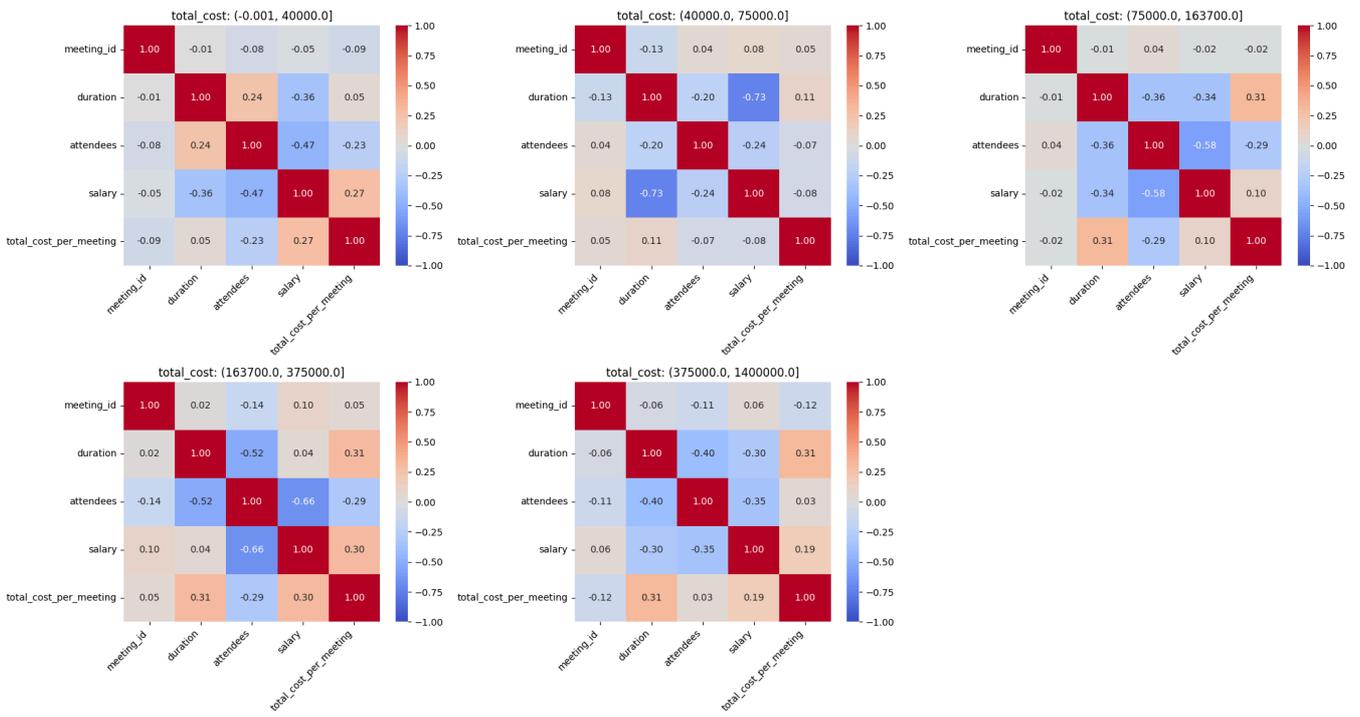
Analyzing correlations by salary

Correlation Heatmaps by salary



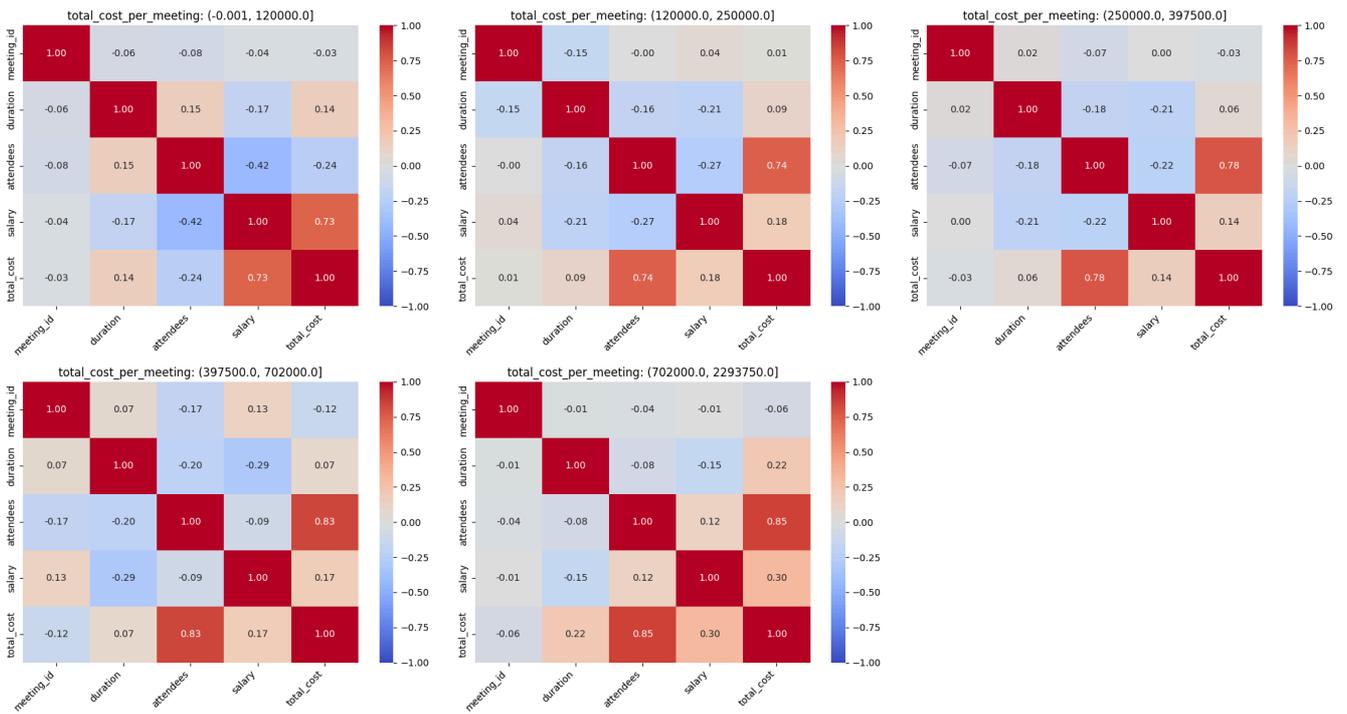
Analyzing correlations by total_cost

Correlation Heatmaps by total_cost



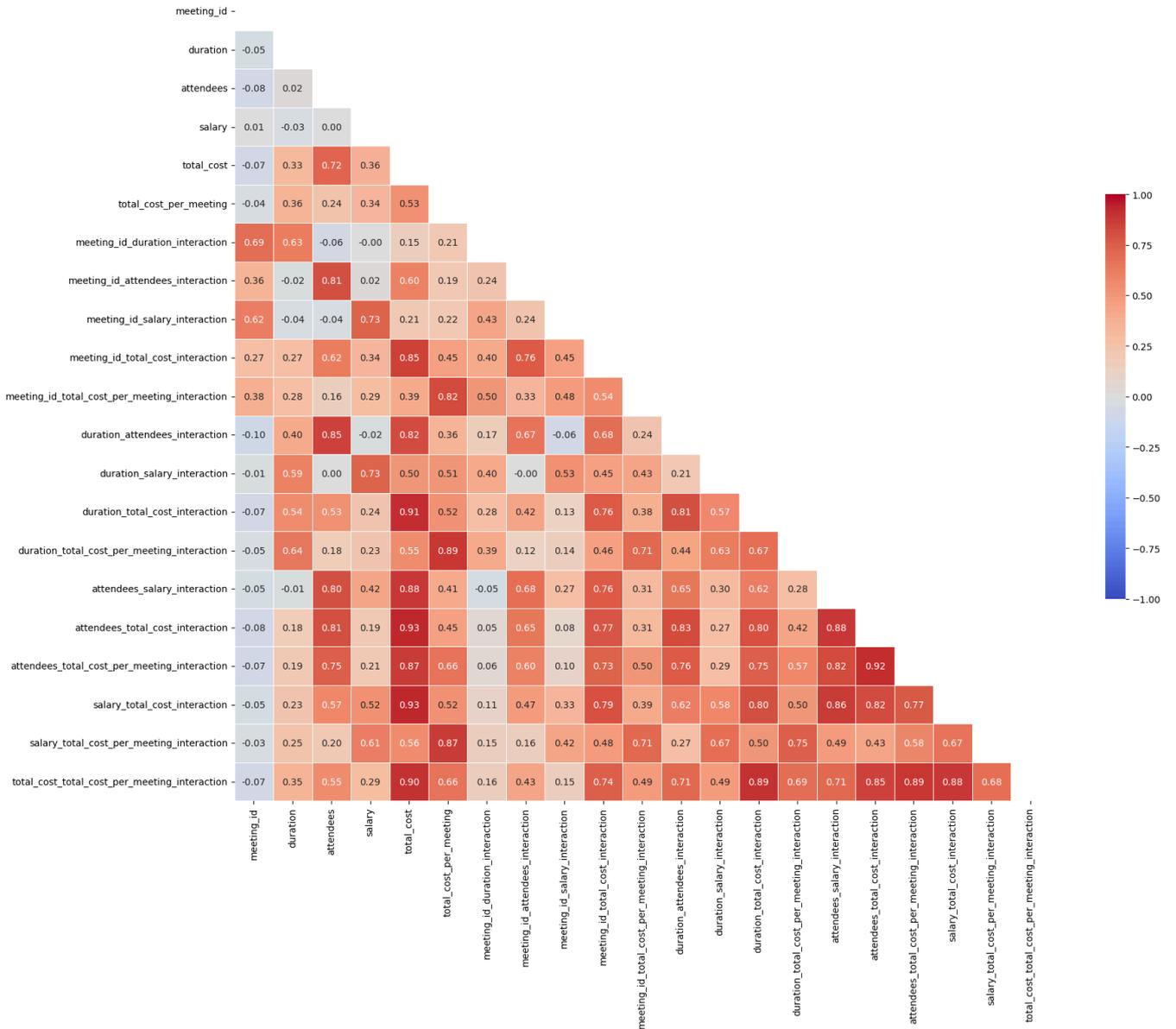
Analyzing correlations by total_cost_per_meeting

Correlation Heatmaps by total_cost_per_meeting



Enhanced Correlation Matrix with Interaction Terms

Enhanced Correlation Matrix with Interaction Terms



Top 10 Highest Correlations:

total_cost	0.928775	attendees_total_cost_interaction
		salary_total_cost_interaction
0.928191		attendees_total_cost_interaction
attendees_total_cost_per_meeting_interaction	0.919330	duration_total_cost_interaction
total_cost	0.913106	total_cost_total_cost_per_meeting_interaction
0.900974		duration_total_cost_interaction
total_cost_total_cost_per_meeting_interaction	0.890676	duration_total_cost_per_meeting_interaction
total_cost_per_meeting	0.886660	attendees_total_cost_per_meeting_interaction
total_cost_total_cost_per_meeting_interaction	0.885984	attendees_salary_interaction
attendees_total_cost_interaction	0.879603	total_cost
attendees_salary_interaction	0.876116	

dtype: float64

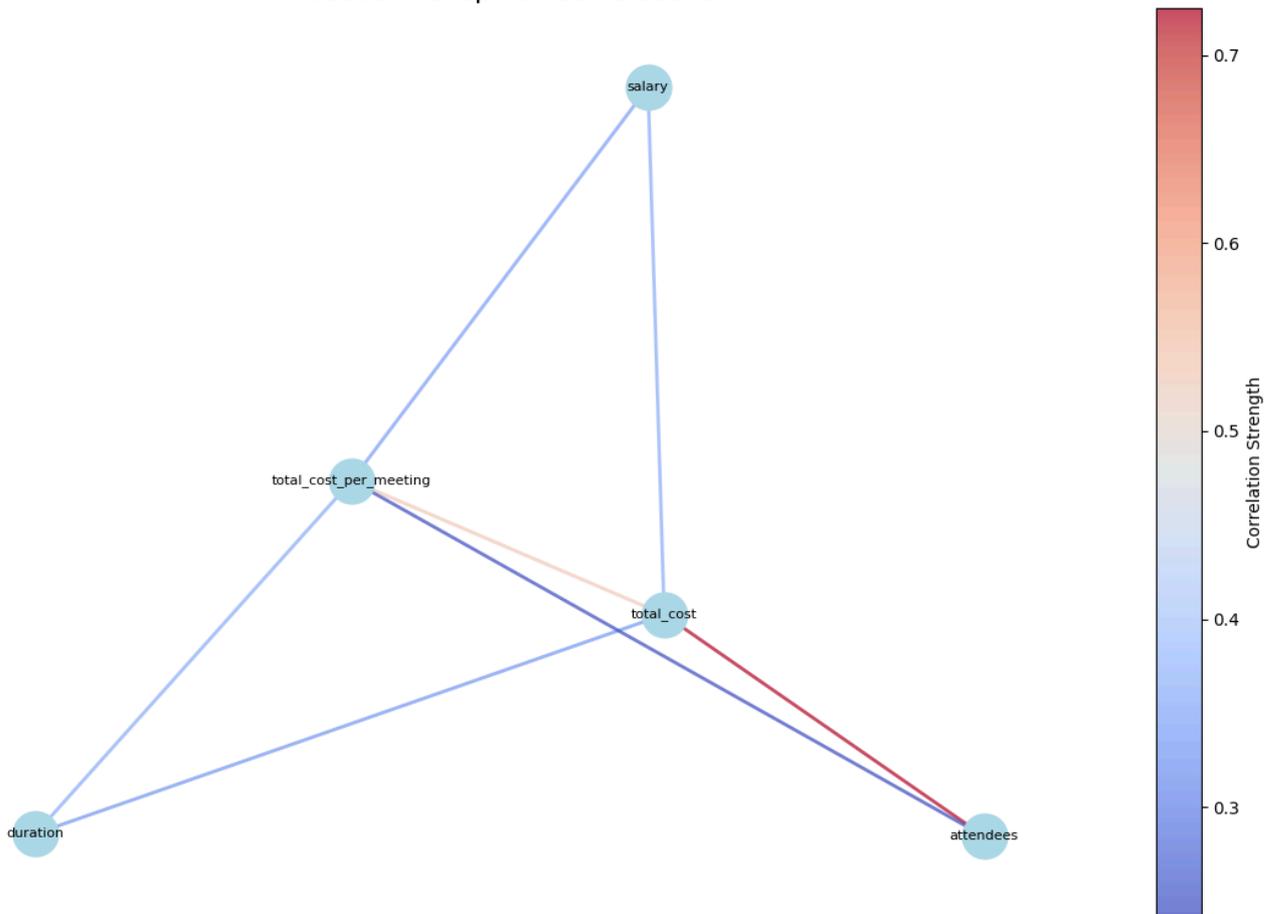
Top 10 Lowest Correlations:

duration	meeting_id_attendees_interaction	-0.019141
meeting_id_attendees_interaction	salary	0.016692
attendees	duration	0.016571
duration	attendees_salary_interaction	-0.014578
meeting_id	duration_salary_interaction	-0.011840
salary	meeting_id	0.005372
	meeting_id_duration_interaction	-0.003347
meeting_id_attendees_interaction	duration_salary_interaction	-0.003261
salary	attendees	0.000209
duration_salary_interaction	attendees	0.000030

dtype: float64

Network Graph of Correlations

Network Graph of Correlations



Top 5 Most Connected Features:

total_cost: 4 connections

total_cost_per_meeting: 4 connections

duration: 2 connections

attendees: 2 connections

salary: 2 connections