STAA57 Group Project

What are the dominant research areas in Ontario's automotive sector, and how do institutions specialize in different fields

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I. Introduction

Research Goal

With the growing popularity of electric vehicles (EVs) and the rapid advancements in artificial intelligence (AI), particularly in the development of autonomous vehicles like Tesla, it is crucial for automotive research facilities in Ontario to focus their efforts and resources on the right areas. This will ensure a significant contribution to the academia world while assisting the government in implementing effective policies and educational institutions better educate, prepare the next generation of labors.

This report aims to examine the automotive research areas prioritizing by Ontario's research facilities and identify specialized facilities in trending fields. It will also explore which areas are attracting the most researchers and which institutions are specializing in particular research domains.

Dataset description

Data source

Our data set was collected by researchers from the Ministry of Economic Development, Job creation and Trade and was published on the Government of Ontario's public database in 2018. A supporting description table of the research tag will be provided in the Appendix for better understanding of the main data set values.

Table 1: Data set Variables

```
## Rows: 545
## Columns: 37
## $ Institution                     <chr> "Brock University", "Carleton Univers~
## $ Researcher.Name                 <chr> "Ahmed, Syed Ejaz", "Ahmadi, Mojtaba"~
## $ Associated.Facilities           <chr> "Centre for Statistical Consulting", ~
## $ Research.Areas                  <chr> "Traffic and road injury prevention."~
## $ Research.Chairs.Grant.Funding   <chr> "", "", "", "", "Canada Research Chai~
## $ Tag.1                           <chr> "Injury Prevention", "Mechatronics", ~
## $ Tag.2                           <chr> "", "Control", "Autonomy and AI", "Ne~
## $ Tag.3                           <chr> "", "", "Sensors", "", "", "", "", ""~
## $ Tag.4                           <chr> "", "", "", "", "", "", "", "", "", "~
## $ Tag.5                           <chr> "", "", "", "", "", "", "", "", "", "~
## $ Alternative.Fuels               <chr> "", "", "", "", "", "", "", "", "", "~
## $ Autonomy.and.AI                 <chr> "", "", "x", "", "", "", "x", "", "",~
## $ Batteries.and.Fuel.Cells        <chr> "", "", "", "", "", "", "", "", "", "~
## $ Biocomposites                   <chr> "", "", "", "", "", "", "", "", "", "~
## $ Coatings.and.Corrosion          <chr> "", "", "", "", "", "", "", "", "", "~
## $ Connected.Vehicles              <chr> "", "", "", "", "", "", "", "", "", "~
## $ Control                         <chr> "", "x", "x", "", "", "", "", "", "",~
## $ Crashworthiness                 <chr> "", "", "", "", "", "", "", "", "", "~
## $ Electronics                     <chr> "", "", "", "", "", "", "", "", "", "~
## $ Forming.and.Joining             <chr> "", "", "", "", "", "", "", "", "", "~
## $ High.Strength.Steel             <chr> "", "", "", "", "", "", "", "", "", "~
## $ Hybrid.and.Electric.Vehicles    <chr> "", "", "", "", "", "", "", "", "", "~
## $ Industrial.Processes            <chr> "", "", "", "", "", "", "", "", "", "~
## $ Injury.Prevention               <chr> "x", "", "", "", "", "", "", "", "", ~
## $ Internal.Combustion.Engines     <chr> "", "", "", "", "", "", "", "", "", "~
## $ Lightweight.Metals              <chr> "", "", "", "", "", "", "", "", "", "~
```

```
## $ Mechatronics                     <chr> "", "x", "", "", "", "", "", "", "", ~
## $ Nanotechnology                   <chr> "", "", "", "", "", "", "", "", "", "~
## $ Networks.and.Security            <chr> "", "", "", "x", "", "x", "", "x", "x~
## $ Noise..Vibration.and.Harshness   <chr> "Err:507", "Err:507", "Err:507", "Err~
## $ Polymers.and.Composite.Materials <chr> "", "", "", "", "", "", "", "", "", "~
## $ Powertrain                       <chr> "", "", "", "", "", "", "", "", "", "~
## $ Sensors                          <chr> "", "", "x", "", "x", "", "x", "", ""~
## $ Software                         <chr> "", "", "", "x", "", "", "", "", "", ~
## $ Stress.and.Fracture              <chr> "", "", "", "", "", "", "", "", "", "~
## $ Transportation.and.Charging      <chr> "", "", "", "", "", "", "", "", "", "~
## $ Vehicle.Design                   <chr> "", "", "", "", "", "", "", "", "", "~
```

Based on a summary of the data set, the `Noise..Viration.and.Harshness` variable was corrupted during the process of uploading or collecting the data and therefore will not be used in our report

Some of the main categorical variables that can help categorizing the data include

- **Researcher.Name**
- **Institution**: Name of university or research center that the researcher works at
- **Research.Areas**: Researcher general field of research
- **Tag**: Researcher's specialized field of research

In addition, the data set also contains research fields as variable (e.g. Alternative Fuels, Autonomy and AI, Vehicle Design) to indicate whether or not a researcher work fall into that field of research

II. Data Overview Analysis (Swaaa)

Descriptive Statistic

Table 2: Summary statistics of research areas (count, frequency)

- Identify the most common research fields

```
research_count = researchers %>% group_by(Tag.1) %>%
  summarise(Count = n()) %>%
  mutate(Frequency = Count / sum(Count)) %>%
  arrange(desc(Count))
top_research_field = research_count$Tag.1[1]
kable(research_count, caption = "Summary statistics of research areas")
```

Table 1: Summary statistics of research areas

Tag.1	Count	Frequency
Networks and security	67	0.1229358
Autonomy and AI	45	0.0825688
Transportation and charging	38	0.0697248
Industrial processes	34	0.0623853
Software	32	0.0587156
Lightweight metals	28	0.0513761

Tag.1	Count	Frequency
Polymers and composite materials	28	0.0513761
Batteries and fuel cells	26	0.0477064
Connected vehicles	23	0.0422018
Forming and joining	21	0.0385321
Vehicle design	21	0.0385321
Alternative fuels	18	0.0330275
Biocomposites	18	0.0330275
Electronics	16	0.0293578
Nanotechnology	16	0.0293578
Control	14	0.0256881
Sensors	14	0.0256881
Coatings and corrosion	13	0.0238532
Internal combustion engines	13	0.0238532
Injury Prevention	12	0.0220183
Hybrid and electric vehicles	9	0.0165138
Mechatronics	7	0.0128440
Noise, vibration and harshness	7	0.0128440
Powertrain	6	0.0110092
Stress and fracture	6	0.0110092
High strength steel	5	0.0091743
Crashworthiness	4	0.0073394
Other	4	0.0073394

- Table 3: Top 5 institutions with the most researchers in the top research field
  - Identify the most dominant institution in a field

```
top_institutions = researchers %>%
  filter(Tag.1 == top_research_field) %>%
  count(Institution, sort = TRUE)%>%
  rename(Reseacher_count = n)%>% head(5)
kable(top_institutions, caption = paste("Top 5 Institutions in", top_research_field))
```

Table 2: Top 5 Institutions in Networks and security

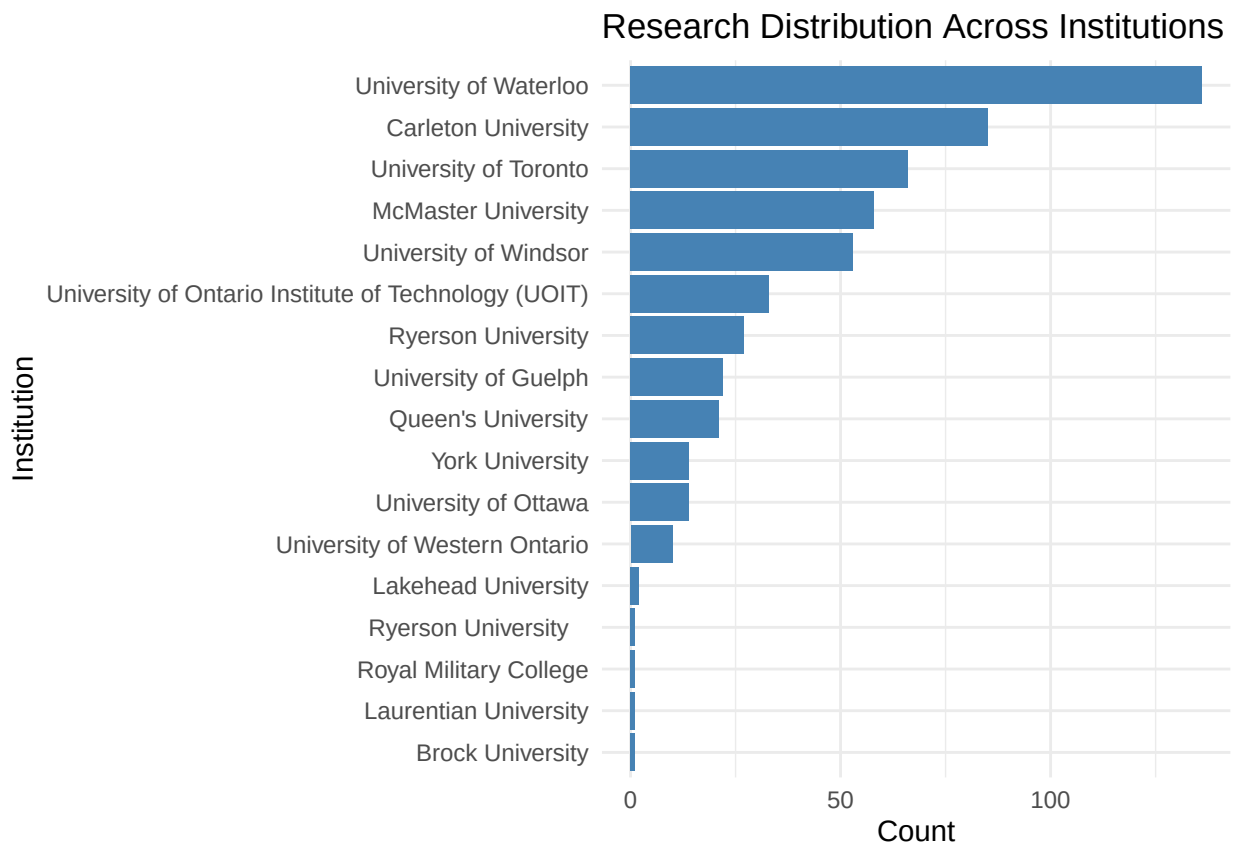
Institution	Reseacher_count
Carleton University	33
University of Waterloo	15
Ryerson University	6
University of Ontario Institute of Technology (UOIT)	3
University of Ottawa	3

Graph & Visualizations

- Bar chart: Research distribution across institution

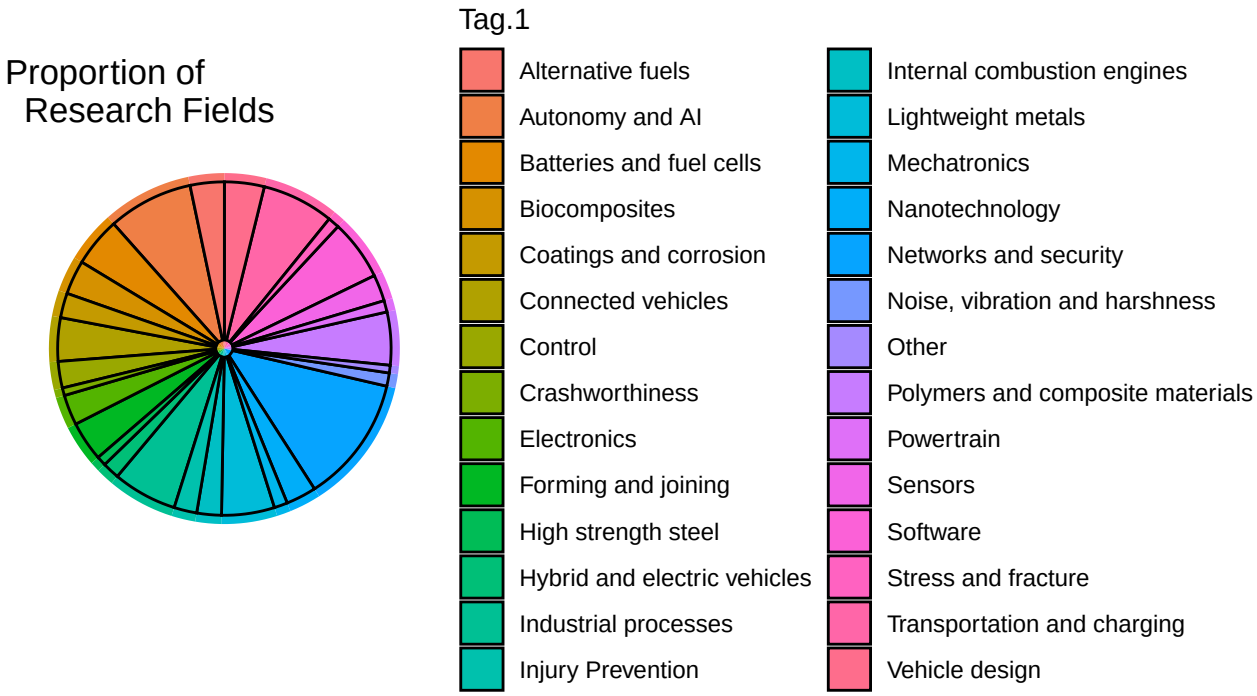
```
ggplot(researchers %>% count(Institution, sort = TRUE), aes(x = reorder(Institution, n), y = n)) +
  geom_bar(stat = "identity", fill = "steelblue") +
  coord_flip() +
```

```
  theme_minimal() +
  labs(title = "Research Distribution Across Institutions",
       x = "Institution", y = "Count")
```



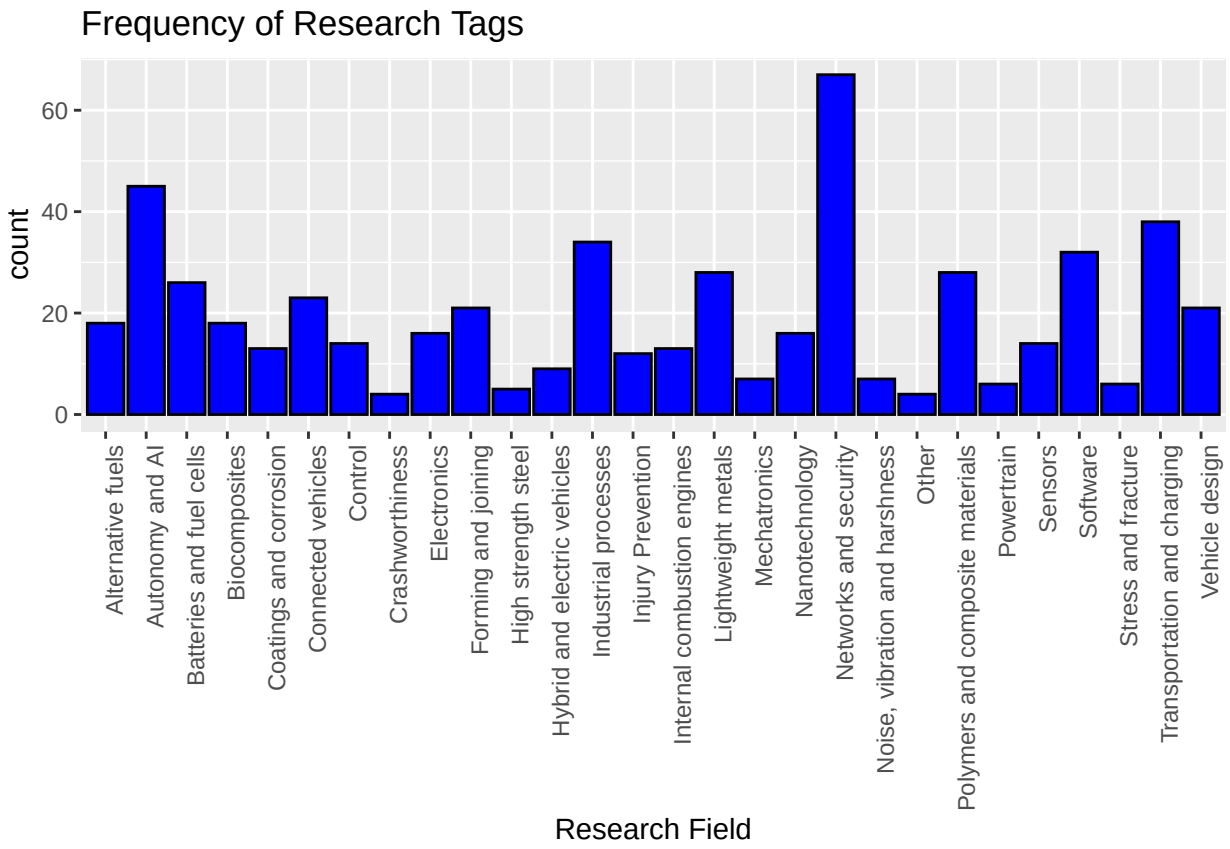
- Pie chart: Proportion of of different research fields

```
ggplot(research_count, aes(x = "", y = Count, fill = Tag.1)) +
 geom_bar(stat = "identity", width = 1) +  geom_col(color = "black") +
 coord_polar("y", start = 0) +
 theme_void() +
 labs(title = "Proportion of
 Research Fields")
```



- History, frequency of research tag

```
ggplot(researchers, aes(x = Tag.1)) +
 geom_bar(fill = "blue", color = "black") +
 theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
 labs(title = "Frequency of Research Tags", x = "Research Field", y = "count")
```



III. Statistical Analysis (rayaanxsyed)

We could pick an area we think that will be most trending (AI for example) and test to see if that is true

- Confidence interval for the average number of researcher
- Hypothesis Testing
- Bootstrapping

IV. Predictive modeling (Regression) (Minh)

1. Logistic Regression Model

1.1. Model explanation

The allocation of research funding plays a pivotal role in identifying key research fields that attract the attention of funding bodies such as Canada Research Chair Program and other major investors. These funding trend are essential indicators of the areas prioritized by the Canadian government and industry stakeholders in the automotive sector. Understanding these trends can provide insight into the direction of research investment, thereby influencing strategic decision in academia and industry alike.

To analyze the factors influencing research funding, we employed a logistic regression model with the primary dependent variable being **is_Funded**. This variable take a value of 1 if the research has secured funding from

the Canada Research Chair program or other similar grants, and 0 otherwise. The independent variables include various research's primary fields as categorized in **Tag.1** column of the data set.

1.2. Result and Key Findings

Table : Regression Model Result Summary

```
##
## Call:
## glm(formula = is_Funded ~ ., family = binomial, data = reg_data)
##
## Coefficients:
##                                   Estimate Std. Error z value Pr(>|z|)
## (Intercept)                       -0.18232    0.60553  -0.301   0.7633
## is_Alternative.Fuels               1.43508    0.82951   1.730   0.0836 .
## is_Autonomy.and.AI                -0.04082    0.67577  -0.060   0.9518
## is_Batteries.and.Fuel.Cells        0.65233    0.72744   0.897   0.3699
## is_Biocomposites                   0.40547    0.76920   0.527   0.5981
## is_Coatings.and.Corrosion          0.02817    0.82231   0.034   0.9727
## is_Connected.Vehicles              0.81093    0.74722   1.085   0.2778
## is_Control                        -1.11696    0.88933  -1.256   0.2091
## is_Crashworthiness                 1.28093    1.30384   0.982   0.3259
## is_Electronics                     0.43364    0.78780   0.550   0.5820
## is_Forming.and.Joining             1.62924    0.82188   1.982   0.0474 *
## is_High.Strength.Steel            -0.22314    1.09545  -0.204   0.8386
## is_Hybrid.and.Electric.Vehicles   -0.04082    0.90370  -0.045   0.9640
## is_Industrial.Processes            0.30010    0.69622   0.431   0.6664
## is_Injury.Prevention               0.51879    0.84233   0.616   0.5380
## is_Internal.Combustion.Engines     1.38629    0.89443   1.550   0.1212
## is_Lightweight.Metals              0.77011    0.72265   1.066   0.2866
## is_Mechatronics                   -0.10536    0.97468  -0.108   0.9139
## is_Nanotechnology                  0.69315    0.79582   0.871   0.3838
## is_Networks.and.Security          -0.53330    0.65905  -0.809   0.4184
## is_Polymers.and.Composite.Materials  0.61764    0.71861   0.859   0.3901
## is_Powertrain                      0.18232    1.01653   0.179   0.8577
## is_Sensors                        -0.40547    0.82327  -0.493   0.6224
## is_Software                       -0.46431    0.71077  -0.653   0.5136
## is_Stress.and.Fracture             0.87547    1.05672   0.828   0.4074
## is_Transportation.and.Charging    -0.35667    0.69265  -0.515   0.6066
## is_Vehicle.Design                  0.47000    0.74907   0.627   0.5304
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##     Null deviance: 755.31  on 544  degrees of freedom
## Residual deviance: 704.64  on 518  degrees of freedom
## AIC: 758.64
##
## Number of Fisher Scoring iterations: 4
```

Table : Regression Model Coefficient

	coefficient	odd
is_Forming.and.Joining	1.6292405	5.10
is_Alternative.Fuels	1.4350845	4.20
is_Internal.Combustion.Engines	1.3862944	4.00
is_Crashworthiness	1.2809338	3.60
is_Stress.and.Fracture	0.8754687	2.40
is_Connected.Vehicles	0.8109302	2.25
is_Lightweight.Metals	0.7701082	2.16
is_Nanotechnology	0.6931472	2.01
is_Batteries.and.Fuel.Cells	0.6523252	1.93
is_Polymers.and.Composite.Materials	0.6176396	1.86
is_Injury.Prevention	0.5187938	1.69
is_Vehicle.Design	0.4700036	1.61
is_Electronics	0.4336360	1.55
is_Biocomposites	0.4054651	1.51
is_Industrial.Processes	0.3001046	1.36
is_Powertrain	0.1823216	1.21
is_Coatings.and.Corrosion	0.0281709	1.03
is_Autonomy.and.AI	-0.0408220	0.97
is_Hybrid.and.Electric.Vehicles	-0.0408220	0.97
is_Mechatronics	-0.1053605	0.91
(Intercept)	-0.1823216	0.84
is_High.Strength.Steel	-0.2231436	0.81
is_Transportation.and.Charging	-0.3566749	0.71
is_Sensors	-0.4054651	0.67
is_Software	-0.4643056	0.63
is_Networks.and.Security	-0.5332985	0.59
is_Control	-1.1169614	0.33

A substantial proportions of the variables int he model exhibit high **p-value** of greater than 5% suggesting that most of the research fields are not statistically significant predictors of whether a research project will be funded. This is likely due to the limited number of observations available int he data set, which may have constrained the model's ability to detect more nuanced relationships. Nevertheless, the model provides valuable insights into general trends in funding allocation by the Canadian government and major investors.

Despite the high **p-value** for many variables, certain research fields stood out interns of their impact on the funding probabilities. Specifically, **Forming and Joining**, **Alternative Fuels**, and **Internal Combustion Engines** demonstrated relatively high log-odds ratios. These fields exhibited 4 to 5 times higher odds of receiving funding than those in other areas. In addition, **is_Forming.and.Joining** and **is_Alternative.Fuels** variables were both statistically significant with p-values of approximately 4% and 10% respectively which further support the findings above.

The significance and high odd of **Forming and Joining** research fields aligns with ongoing research efforts to improve manufacturing process which are crucial for the production of commercial vehicles. While the corresponding number for **Alternative Fuels** underscores the importance of environmentally sustainable technologies, which are increasingly emphasized in both governmental policy and industry innovation.

2) Cross validation

To further evaluate the robustness of the logistic regression model, we conducted a k-fold cross-validation with four folds.

The results are as follow:

```
## AUC score: 0.6114087 0.6021347 0.4954117 0.6215278
```

```
## Average AUC score: 0.5826207
```

The model predictive performance assessed using the AUC score. The resulting average AUC score was 0.609, which suggests a fair to somewhat weak performance. This moderate performance is likely attributed to the large number of non-significant variables in the model, reflecting the challenges posed by limited data. Nevertheless, the model remains useful for identifying general funding trends rather than providing highly accurate predictions for individual research projects. It serves its purpose of illustrating that dominant research areas that are attracting funding, which is the primary objective of the analysis.

V. Summary

In conclusion...

The logistic regression analysis provides valuable insights into the research funding landscape within the Canadian automotive sector. The significant fields of **Forming and Joining** and **Alternative Fuels**, along with the notable importance of **Internal Combustion Engines**, underscore the ongoing focus on traditional automotive manufacturing processes and sustainable technologies. These finding aligns with the broader trends observed in the automotive industry, where there is increasing attention to environmentally friend alternatives to conventional fuels and more efficient manufacturing practices.

However, as the market of Autonomous Vehicles and Electric Vehicles grows, we believe that more resources should be reallocate to support researches in these emerging areas. Fields such as **Autonomy and AI**, **Batteries and Fuel Cells**, and **Hybrid and Electric Vehicles** are expected to play a pivotal role in shaping the future of automotive industry. # Appendix