



CAR DAMAGE DETECTION

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TABLE OF CONTENTS

Table of Contents	ii
List of Abbreviations	iv
1 Realistic Constrains	1
1.1 Data Availability	1
1.2 Time Constraints	1
1.3 Budget Limitations	2
2 Social, Environmental and Economic Impact	3
2.1 Social Impact	3
2.1.1 Streamlined Insurance Claims Processing	3
2.1.2 Job Creation and Education	3
2.1.3 Enhanced Safety and Awareness	3
2.2 Environmental Impact	3
2.2.1 Resource Optimization	3
2.2.2 Sustainable Practices	4
2.3 Economic Impact	4
2.3.1 Cost Savings	4
2.3.2 Revolutionizing the Car Insurance Industry	4
2.3.3 New Market Opportunities	4
2.3.4 Economic Stability	5
2.4 Ethical Considerations and Future Directions	5
2.4.1 Ethical Application	5
2.4.2 Continuous Improvement	5
2.4.3 Regulatory Compliance	5
3 Cost Analysis	6
3.1 Salaries	6
3.2 Software	6
3.3 Equipment	6
3.4 Possible Future Costs	7
3.4.1 Data Acquisition	7
3.4.2 Upgraded Software and Tools	7
3.4.3 Hardware Upgrades	7
3.4.4 Miscellaneous Costs	7

4	Risk Analysis	8
4.1	Data Privacy Risk	8
4.2	Model Accuracy Risk	8
4.3	System Security Risk	9
4.4	Operational Risk	9
4.5	Reputation Risk	10
5	Standards and Ethical Guidelines	11
5.1	IEEE Standards	11
5.2	European Commission AI Ethics Guidelines	12
6	Conclusion	14
	References	15

LIST OF ABBREVIATIONS

AI	Artificial Intelligence
GPU	Graphics Processing Unit
IEEE	Institute of Electrical and Electronics Engineers
R-CNN	Region-based Convolutional Neural Network
TL	Turkish Lira
USD	United States Dollar

1 Realistic Constrains

1.1 Data Availability

- **Quality of Data:** For the model to function in an accurate manner, it will require highly detailed images and perfectly annotated data. On the other hand, if they receive just any poor quality pictures with insufficient annotations this can lower the effectiveness of a model.
- **Quantity of Data:** The model generalizes better with a bigger data set. Lack of enough data might cause overfitting meaning that a model does well with certain values during training while failing on new ones.
- **Data Distribution:** Ensure data is evenly distributed across different damage types, vehicle models, and environmental conditions. Uneven distribution can bias the model equipment.

Recommendations:

- **Data Augmentation:** Rotation, flipping, scaling, changes in color hail to increase dataset size artificially.
- **Crowdsourcing:** Amazon Mechanical Turk is a good platform for data annotation.
- **Public Datasets:** Reuse or utilize some public data sets relevant to the task of Vehicle Damage Detection to enrich your data.

1.2 Time Constraints

- **Training Time:** Google Colab's free tier has usage limits, leading to session interruptions.
- **Compute Resources:** Limited access to high-performance GPUs can slow down the training process.

Recommendations:

- **Checkpointing:** Save the model's progress and then resume from the last saved state to continue training where it left off in the event of an interruption.
- **Smaller Batches:** Train with smaller batches if GPU memory is a constraint.
- **Alternative Platforms:** Explore other platforms like Kaggle Kernels, AWS Educate, or Azure for Students for additional compute resources.

1.3 Budget Limitations

- **Data Acquisition Costs:** Financial constraints may limit purchasing high-quality datasets or collecting new data.
- **Equipment Costs:** High-end GPUs, storage solutions, and other hardware may be beyond budget. process.

Recommendations:

- **Open-Source Tools:** Utilize free, portable tools and libraries instead of purchasing pricey licenses for software.
- **University Resources:** Leverage institutional resources like lab computers, cloud credits, or datasets.
- **Grants and Sponsorships:** Seek academic grants or corporate sponsorships to support your project.

2 Social, Environmental and Economic Impact

2.1 Social Impact

2.1.1 Streamlined Insurance Claims Processing

- **Efficiency:** The process of insurances claims can be expedited by using artificial intelligence which automates the detection of car damages. As a result, settlements are reached faster and customer satisfaction is improved.
- **Minimized Disputes:** With accurate and objective damage assessments, disputes between insurance companies and claimants can be minimized, fostering trust and transparency in the process.

2.1.2 Job Creation and Education

- **New Job Opportunities:** The use of AI for car damage detection can create new jobs in the area like AI developers, data annotators or model maintainers.
- **Educational Advancement:** Universities and training institutions can incorporate such advanced AI applications into their curriculum, preparing students for future job markets and encouraging research in AI.

2.1.3 Enhanced Safety and Awareness

- **Preventive Measures:** Accurate damage detection can help in identifying issues that need immediate attention, potentially preventing accidents.
- **Public Awareness:** Increased use of AI in everyday applications can raise public awareness about the benefits and capabilities of AI technologies.

2.2 Environmental Impact

2.2.1 Resource Optimization

- **Efficient Repairs:** By accurately identifying the extent of damage, the AI system ensures that only necessary repairs are made, reducing waste and conserving resources.

- **Reduced Carbon Footprint:** Efficient repair processes can lead to fewer unnecessary replacements and less production of new parts, thereby reducing the overall carbon footprint of the automotive repair industry.

2.2.2 Sustainable Practices

- **Promoting Sustainability:** The adoption of AI for damage detection can encourage the automotive industry to adopt more sustainable practices, such as recycling parts and reducing material waste.
- **Conservation of Resources:** Optimizing repair processes helps in conserving raw materials and reducing the environmental impact of manufacturing and transportation of new parts.

2.3 Economic Impact

2.3.1 Cost Savings

- **Avoiding Unnecessary Repairs:** Accurate damage detection helps in identifying the exact repairs needed, avoiding unnecessary expenses and reducing overall repair costs.
- **Efficient Resource Allocation:** Insurance companies can allocate resources more efficiently, reducing overhead costs associated with lengthy claims processing.

2.3.2 Revolutionizing the Car Insurance Industry

- **Faster Claims Processing:** When AI is integrated into the system of claim processing, it has the potential to remarkably shorten claim cycles, thus making payments faster and greatly improving cash flow for both insurers and their clients.
- **Accurate Assessments:** The ability to identify a more precise extent of loss will assist in reducing fraud incidence and underinsurance cases.

2.3.3 New Market Opportunities

- **AI-Based Solutions:** Success in applying artificial intelligence in car damage detection could result in new opportunities for similar AI based solutions in other industries such as property damage assessments, healthcare diagnostics etc.

- **Business Growth:** Companies that develop and offer AI based solutions can experience immense growth that leads to increased investment in AI technology advancement.

2.3.4 Economic Stability

- **Reduced Litigation Costs:** With accurate appraisals that minimize disputes, insurance firms are able to diminish litigation expenses hence enhancing economic stability within this sector.
- **Job Market Expansion:** The spread of applications using AI can create jobs hence leading to economic development.

2.4 Ethical Considerations and Future Directions

2.4.1 Ethical Application

- **Bias and Fairness:** To avoid bias and ensure fairness across different demographics and vehicle types, make sure that the AI model is trained on various datasets.
- **Transparency:** Transparency must be maintained so that AI model works, and decisions are made transparently to allow users and stakeholders trust their thoughts.

2.4.2 Continuous Improvement

- **Refinement of Models:** The AI models continuously need to be refined and updated for accuracy and reliability improvement.
- **User Feedback:** Incorporate user feedback to enhance the system and address any emerging issues.

2.4.3 Regulatory Compliance

- **Adherence to Regulations:** To ensure ethical and legal integrity, AI use of car damage detection should comply with applicable regulations and industry standards.

3 Cost Analysis

In this section, we will talk about the cost spent to carry out this study and the costs that may arise in possible cases.

3.1 Salaries

Artificial Intelligence Engineers:

- **Monthly salary per engineer:** 17,000 TL
- **Number of engineers:** 3
- **Project duration:** 8 months
- **Total = 3 engineers \times 17,000 TL/month \times 8 months = 408,000 TL**

3.2 Software

Tools Used:

- **Roboflow** Free version used for dataset management and preprocessing.
- **PyCharm:** Free version used for development and coding.
- **Pro version of Google Colab:** \$10 per month

Total Software Cost:

- **Total = \$10 USD/month \times 8 months = \$80 USD \approx 2,400 TL (assuming 1 USD = 30 TL)**

3.3 Equipment

Hardware:

- Utilized Google Colab's cloud GPUs, thus no additional hardware costs were incurred.

3.4 Possible Future Costs

3.4.1 Data Acquisition

- The model may require additional datasets to improve its accuracy and robustness.
- Costs vary depending on the source and quality of data.

3.4.2 Upgraded Software and Tools

- In case the project expands, more advanced tools or additional software licenses might be needed.
- Potential use of commercial tools for data labeling, storage, or additional cloud services..

3.4.3 Hardware Upgrades

- Investment in high-performance GPUs or dedicated cloud services may be required if there is an increase in computational needs.
- High-performance hardware setup reference costs could differ widely.

3.4.4 Miscellaneous Costs

- Costs for workshops, training sessions, or professional development for team members.
- Travel and communication expenses for collaboration with external partners or attending conferences.

4 Risk Analysis

4.1 Data Privacy Risk

Description: The risk of violating user privacy by mishandling user data, such as images of vehicles uploaded to your web page.

Potential Impact: The consequences of mishandling user data can include legal repercussions, loss of user trust, and potential financial penalties.

Solution:

- **Robust Data Handling and Privacy Policies:** Develop and implement comprehensive data privacy policies.
- **Data Anonymization and Encryption:** Ensure that all user data is anonymized and encrypted to protect user identities.
- **Compliance with Regulations:** Adhere to local and international data protection regulations such as GDPR, CCPA, etc.
- **Regular Audits:** Conduct regular audits to ensure compliance and address any vulnerabilities.

4.2 Model Accuracy Risk

Description: The risk that the model may not accurately identify damage in certain scenarios, leading to incorrect results.

Potential Impact: Incorrect assessments, user dissatisfaction, and potential financial losses.

Solution:

- **Regular Updates and Training:** Continuously update and retrain the model with diverse and high-quality data.
- **Feedback System:** Implement a feedback mechanism where users can report inaccuracies, which can be used to improve the model.
- **Validation and Testing:** Regularly validate and test the model on different datasets to ensure its accuracy and reliability.

4.3 System Security Risk

Description: The risk of cyber-attacks that could compromise the system and the data it holds.

Potential Impact: Data breaches, loss of sensitive information, and operational disruption.

Solution:

- **Strong Security Measures:** Implement robust security protocols, including firewalls and intrusion detection systems.
- **Regular Security Audits:** Conduct regular security audits to identify and mitigate vulnerabilities.
- **Data Encryption:** Ensure that all data, both in transit and at rest, is encrypted.
- **Access Control:** Implement strict access control measures to limit access to sensitive data.

4.4 Operational Risk

Description: The risk of system failure or downtime, which could disrupt the service.

Potential Impact: Service disruption, loss of user trust, and financial losses.

Solution:

- **Reliable Backup and Recovery Procedures:** Establish and regularly test backup and recovery procedures to ensure quick restoration in case of failures.
- **Redundancy:** Implement redundancy in system architecture to minimize downtime.
- **Monitoring and Maintenance:** Regularly monitor system performance and conduct maintenance to prevent unexpected failures.

4.5 Reputation Risk

Description: The risk of damage to your reputation if the system makes a mistake or is involved in a controversy.

Potential Impact: Loss of user trust, negative publicity, and financial impact.

Solution:

- **Transparency:** Clearly communicate the system's capabilities and limitations to users.
- **Prompt Response:** Address any issues or controversies promptly and professionally.
- **Quality Assurance:** Maintain high standards of quality assurance to minimize errors and build user trust.

5 Standards and Ethical Guidelines

This section examines the standards and ethical guidelines related to the development of car damage detection systems using the Detectron2 Mask R-CNN model.

5.1 IEEE Standards

IEEE standards ensure the quality, reliability, and safety of software and system engineering projects. These standards are crucial when developing a car damage detection system using the Detectron2 Mask R-CNN model.

1. IEEE 12207:2017 - Software Life Cycle Processes

Description: This standard provides a common framework for managing and improving software development processes.

Application:

- **Development Stages:** Ensure that all stages of the model development, from requirements gathering to deployment and maintenance, adhere to the IEEE 12207:2017 standards.
- **Process Improvement:** Continuously monitor and improve development processes to maintain high quality and efficiency.

2. IEEE 1012-2016 - Verification and Validation Description:

This standard offers a rigorous methodology for ensuring the accuracy and reliability of software products.

Application:

- **Model Verification:** Implement thorough verification processes to ensure that the Detectron2 Mask R-CNN model meets its specifications.
- **Model Validation:** Conduct extensive validation to confirm that the model performs accurately and reliably in real-world scenarios.

3. IEEE 7001-2021 - Transparency of Autonomous Systems Description:

This standard ensures transparency in the decision-making processes of AI models.

Application:

- **Decision Transparency:** Make the decision-making process of the Detectron2 Mask R-CNN model transparent to users.

- **User Trust:** Provide clear and understandable explanations of how the model operates to build user trust.

5.2 European Commission AI Ethics Guidelines

These guidelines ensure that AI systems are developed in an ethical and trustworthy manner.

1. Respect for Human Autonomy

Description: AI should support human decision-making and leave final decisions to humans.

Application:

- **Decision Support:** Design the car damage detection system to assist users in making informed decisions, without removing human oversight.
- **Final Authority:** Ensure that the final decision-making power remains with the human user.

2. Prevention of Harm

Description: AI systems should avoid making incorrect assessments that could harm users.

Application:

- **Model Accuracy:** Regularly update and improve the model to minimize errors and avoid incorrect assessments.
- **Risk Mitigation:** Implement safeguards to prevent the model from causing harm due to incorrect predictions.

3. Fairness

Description: AI systems should produce fair and unbiased results for all users.

Application:

- **Bias Mitigation:** Continuously monitor the model for biases and take corrective actions to ensure fairness.
- **Inclusive Design:** Ensure that the model is trained on diverse data to produce equitable results across different user groups.

4. **Explicability Description:** The model's predictions should be understandable and explainable to users..

Application:

- **Transparent Algorithms:** Ensure that the algorithms used in the car damage detection system are transparent. Users should be able to understand how the system arrived at its conclusions.
- **User-Friendly Explanations:** Provide clear and concise explanations for the model's predictions. For instance, if the AI detects a certain level of damage, it should explain the criteria used for this assessment in simple terms.
- **Documentation and Training:** Offer comprehensive documentation and training to users so they can understand and trust the AI system. This could include user manuals, FAQs, and training sessions.

These standards and ethical guidelines ensure that your car damage detection system is developed and used in a reliable, safe, and ethical manner.

6 Conclusion

In summary, the development of the Car Damage Detection system aimed to provide an efficient and accurate solution for identifying and assessing car damage using deep learning techniques. Utilizing the Detectron2 framework and Streamlit, the system effectively delivers a user-friendly experience for diverse users, including insurance companies, repair shops, and vehicle owners.

Throughout development, we navigated several constraints. Economically, we managed costs by leveraging open-source tools and utilizing university resources. Technologically, we ensured high performance by using cloud-based GPUs and implementing robust data handling practices. User constraints were addressed with an intuitive web application interface and real-time feedback. Regulatory compliance with GDPR and other relevant standards was ensured to protect user data. Ethically, we ensured fairness and transparency in the AI model's operation.

The Car Damage Detection system enhances the automotive and insurance industries by automating damage assessment processes. Socially, it streamlines insurance claims and enhances safety by providing accurate damage assessments. Environmentally, it promotes resource optimization and sustainable practices. Economically, the system generates savings by avoiding unnecessary repairs and revolutionizing the car insurance industry.

Cost analysis included personnel, software licenses, and potential future expenses for data acquisition, upgraded tools, and hardware. Effective cost management was achieved by leveraging open-source tools and seeking grants and sponsorships.

Adhering to IEEE and European Commission AI ethics guidelines ensured quality, safety, and reliability. Compliance with GDPR guaranteed user privacy. Potential risks such as data privacy, model accuracy, and system security were addressed through comprehensive testing, continuous improvement, and strong security measures.

In conclusion, the Car Damage Detection system is a comprehensive, user-friendly application designed to automate vehicle damage assessment effectively. It will continue to evolve with future updates, enhancing its features and incorporating new technologies, positioning itself as a leading tool in the automotive industry.

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