### Data Science Project Checklist

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This checklist is a print-friendly, interactable translation of <u>DataCamp's Data Science Project Checklist</u>. DataCamp describes the checklist as "summarizing data science project management best practices..." and "...combines CRISP-DM project management principles with those of the Agile and Scrum software development frameworks."

The following characteristics should underpin any data science project:

- **Measurable**: Is the success of a project and its impact on the business quantifiable?
- **Reliable**: What proportion of projects achieved their goals?
- **Scalable**: Can the throughput of projects be increase without significantly degrading reliability?

Additionally, the following principles should be abided by:

- Iterative
- Reusable and recyclable
- Reproducible

The framework followed will be:

Context setting & ideation	Data collection & exploration	Modeling & testing	Deploy & user testing	Monitoring
Project Name:				
Project Start Date:				
Notes:				

## Context Setting & Ideation

	$\operatorname{Task}$	Notes
	Identify the business	
	problem being solved.	
	Clarify why the project is	
	being undertaken as	
	unambiguously as possible.	
	Identify stakeholders.	
П	Roles may include project	
	manager, data scientist,	
	account manager, etc.	
	Review prior work.	
	Review existing projects	
	that covered similar	
	ground. What were key	
	outcomes? Can anything be	
	reused? What mistakes	
	should be avoided?	
	Determine key performance	
	indicators to measure	
	success.	
	Metrics should use SMART	
	criteria.	
	Determine the scope.	
	What are the deliverables?	
	What are the reqs. for the	
	deliverables? What will not	
	be included?	
	Write a project plan.	
	Set milestones for	
	intermediate steps in the	
	project. Decide on timeline	Do external to this document.
	for each milestone. Write a	
	short description of each	
	step.	

	Estimate the impact of the	
	project.	
	Quantify the benefit to the	
	organization if the goals	
	are reached. List	
	qualitative benefits.	
	Estimate the effort of the	
	project.	
	How much will the project	
	cost? How much time will	
	the project take? What	
	resources will it need?	
	Estimate the project risks.	
	List all the risks to the	
	project. Calculate the risk	
	impact the probability of it	
	occurring times severity of	
	it occurring.	
	Decide whether or not to	
	proceed with the project.	
	Decide to proceed now, put	
	on hold, or cancel.	
	Determine the	
	responsibility of each	
	stakeholder.	
	Use the RACI model for	
	each task.	
	Determine a	
	communication strategy.	
	How will you keep in	
	touch? Meeting cadence?	
	Identify data sources.	
	Do you have access? Where	
	is it stored? What form is	
	it in? Etc.	
	Anticipate regulatory	
	needs.	

#### Context Setting & Ideation

	Will anything be audited?	
	Can data sources be legally	
	used?	
	Decide on a technology	
	stack.	
	Agree on tools for storing,	
	processing, and modeling	
	the data.	
	Write a project charter.	
	Summarize what was	Do external to this document.
	decided in a short	Do external to this document.
	document.	

# Data Collection & Exploration

$\operatorname{Task}$	Notes
Give data scientists access	
to all datasets.	
Organize appropriate	
permissions. Purchase any	
commercial datasets or use	
synthetic data with similar	
properties.	
Ingest the data.	
For each data source, move	
it to the analytics	
environment.	
Explore the data.	
Visualize distribution of	
each variable with	
histogram or bar plot.	
Quantify missing values for	
each variable. Visualize	
relationship between	
features and target variable	
with scatter, histogram,	
box, or heatmap plots.	
Determine key performance	
indicators to measure	
success.	
For each dataset: provide a	
summary, describe high-	
level data quality issues,	
describe the quality of	
target variable, describe	
quality of each feature,	
describe relationship	
between each feature and	
the target variable.	

#### Data Collection & Exploration

	Decide whether or not to	
	proceed with the project.	
	Continue, pause, or cancel.	
	Build a data pipeline.	
	Pipeline should automate	
	the ingestion and cleaning	
	process. Should run as	
	batch or stream job.	
	Document the data	
	pipeline.	
	Draw diagram of steps in	Do external to this document.
	the data pipeline and their	Do externar to this document.
	dependences. Describe	
	what happens in each step.	

### Modeling & Testing

$\operatorname{Task}$	Notes
	Modeling
Generate a hypothesis.	
Does the hypothesis make	
sense in the business	
domain? Can you measure	
the outcome? Do you have	
sufficient data to be	
significant? Watch for	
biases.	
Split your data into	
training and testing sets.	
Do this before engineering	
features to ensure no data	
leakage.	
Engineer features.	
Center/scale numeric	
values, create categorical	
variables (incl. binning),	
apply Box-Cox or Yeo-	
Johnston transformations	
to numeric variables to give	
normal dist., combine	
rare/related categories of	
categorical variables,	
extract or combine parts of	
datetimes, create new	
variables from summary	
statistics, extract	
quantitative metrics from	
text and other unstructured data.	
 Fit the model/run an	
experiment.	
Fit simplest first then	
increase complexity.	

Report on the results.	
Regularly provide feedback	
to stakeholders, adjust	
language accordingly,	
report failures as well as	
successes.	
	Testing
Create a test suite.	
Unit tests or back test that	
run automatically to check	
your model or experiment's	
performance.	
Validate the business	
impact.	
Discuss the impact with	
business stakeholders.	
Validate the technical	
approach.	
Check assumptions of your	
model to ensure they're	
valid. See if results are	
sensitive to sampled data.	
Are hyperparameters	
suitable? Is the model	
reproducible?	
Validate the deployability.	
Can all input values or use	
cases be handled? Are all	
required data sources	
available in production?	
Can model fail gracefully?	
Can predictions be made	
fast enough?	
 Preserve null results.	
Record anything that won't	
make it to production.	

## Deployment & User Testing

	$\operatorname{Task}$	Notes
		Deployment
	Develop a data pipeline.	
	Set up DAG for all the	
	data sources to production	
	environment.	
	Develop a model pipeline.	
	Provide an API to your	
	model that can be accessed	
	by dashboard/website/etc.	
	Design a monitoring plan.	
	Determine metrics to be	
	tracked, including	
	performance and safety	
	metrics. Determine limits	
	for acceptable ranges for	
	those metrics. Decide how	
	you wish to be alerted if	
	metrics go out of range.	
	Roll out via A/B test.	
	Provide the new	
	feature/model to random	
	sample of users. Monitor	
	chosen metrics. Declare	
	winner when there is	
	statistical significance.	
	Analyze and report on A/B	
	test results.	
	Compare the metrics	
	chosen to track for each	
	group. Report results even	
	if test was not a success.	
	Roll out to most or all	
	users.	
	If successful, rollout to all	
	but small holdout group.	

	User Testing
Write an exit report.  Summarize status of project and what you learned. Provide overview of the project. Summarize business problem you tried to solve. Describe the data sources and how they were processed. Describe modeling techniques used and their validation. Summarize solution architecture. Outline benefits from the project for company and customer. Describe any learnings around project execution, data science, business domain, and the product. Outline next steps.	Do external to this document.
Get customer feedback.  Conduct surveys and user interviews. Monitor reviews, ratings, and social media.	

### Monitoring

	Task	Notes	
	Develop monitoring		
	pipeline.		
	Set up pipeline to		
	automatically track the		
	performance and safety		
	metrics defined in the		
	monitoring plan.		
	Create dashboards.		
	Create dashboards to track		
	the changes of metrics over		
	time.		
	Set up alerts.		
	Set up notifications via		
	email, Slack, etc. when		
	metrics fall out of the		
	acceptable range.		
Pro	Project End Date:		
	Time Elapsed:		