

Edgar

Micah Blake McCurdy
micah@hockeyviz.com

Rochester NY
RIT Hockey Analytics Conference
October 20, 2017

Models, As Abstractly As Possible

A *model* is a way to gather some important aspects of an interesting thing, so that we can benefit.

Friendly, Familiar Models

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- ▶ Models from physics:

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 - ▶ Hockey!

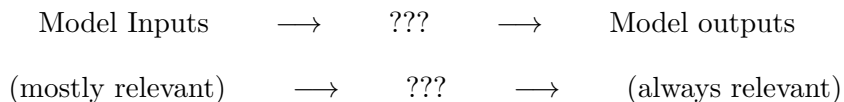
What Makes A Model Good?

- ▶ Accuracy
- ▶ Efficiency
- ▶ Interpretability

Relevance

Model Inputs \longrightarrow ??? \longrightarrow Model outputs

Relevance



Model Classification

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- ▶ When the inputs are combined by some encoding of the mechanics of the thing being modelled, that makes the model *phenomenological*.
 - ▶ When the combining is done systematically with oversight, that makes the model *scientific*.

Edgar

I made what I think is a “scientific” model in this sense and I called it: Edgar

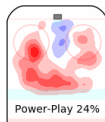
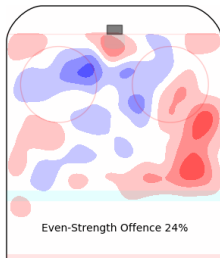
Which aspects are important?

- ▶ Unblocked shot rates and their locations.
- ▶ Penalty rates
- ▶ Shooting talent
- ▶ Goaltending talent
- ▶ Who takes the shots
- ▶ Rest

I estimate them all with statistics.

Isolating Player Ability

Player Isolate, Erik Karlsson

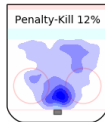
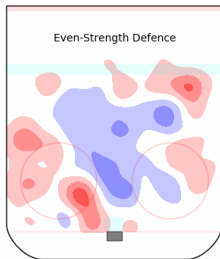


Relative to League

Shooting: +0.8%

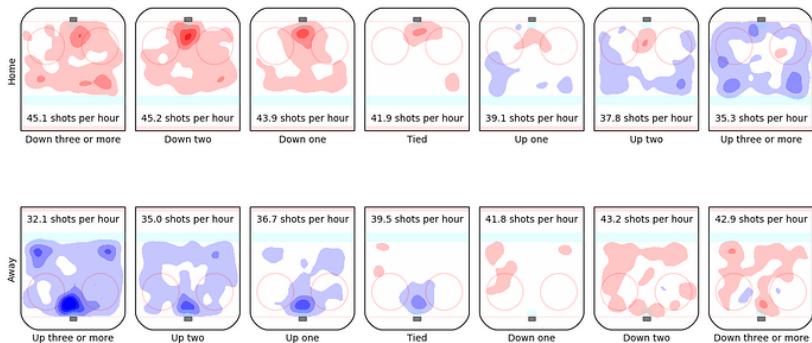
Penalties Drawn: -30%

Penalties Taken: -27%



Isolating Player Ability: Shot Maps

Shot rate maps are adjusted for score



Some day: adjusted for teammates (and then maybe zone usage and competition)

Shooting and Goaltending Abilities

All measured *relative to where shots are taken*.

Shooters

Excellent shooters:

Jake Guentzel	+6.5%
Patrik Laine	+6.2%
Sven Baertschi	+5.1%
Jannik Hansen	+4.5%

(Also Ho-Sang, Malkin, Barkov, Gourde, Athanasiou, Oshie)

Regressed 2/3 of the way to the mean

Goalies

Excellent goalies:

Brayden Holtby	+1.3%
Matt Murray	+1.2%
Carey Price	+1.2%
Henrik Lundqvist	+1.2%

(Also Smith, Grubauer, Crawford, Gibson, Reimer, Bobrovsky)

Regressed 1/2 of the way to the mean

Additional Sneakiness

Penalty Rates and Shot Propensities: Untouched! (for now)

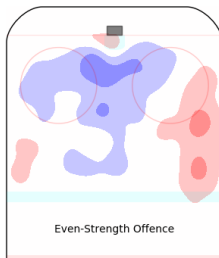
Rookies treated as league average, except for a chosen few.
(estimated by Hannah Stuart)

Some ad-hoc regression for people with very little relevant icetime.

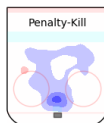
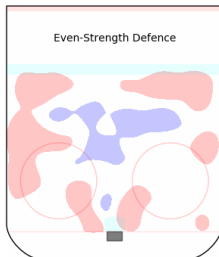
Isolating Team Abilities

Estimate
team
rosters

Team Isolate, OTT



Relative to League
Shooting: +0.29%
Goaltending: +0.21%
Penalties Drawn: -7%
Penalties Taken: -3%



Adjust for rest

Simulation Mechanism

Model shots and penalties as Poisson processes with the measured rates.

For every shot taken:

- ▶ Choose a location
- ▶ Choose a shooter
- ▶ Adjust for shooting talent
- ▶ Adjust for goalie
- ▶ See if it's a goal

And so on, for sixty or perhaps sixty-five minutes.

Information in Excess of Guessing

My preferred measure of accuracy for single games:

$$100 \log_2 2p$$

where p is the probability for the outcome that happened.

Really just log-loss, scaled onto 0 (guessing) and 100 (perfection).

Results From 2016-2017

Creator	Model	Information In Excess of Guessing (per game)
Perry	Salad	5.03
Nandakumar	Feline Frenzy	4.65
Luszczyszyn	Preszczyszyn	4.36
M.	Edgar	4.13
Sprigings	DTMAH	4.10
M.	Cordelia	2.11

Thanks!

