

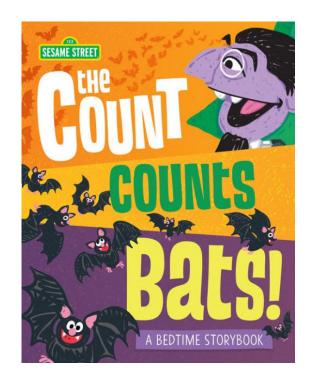
Goals of wildlife management

- 1. make them increase
- 2. make them decrease
- 3. keep them stable
- 4. do nothing but keep an eye on them

What do we need to know!?

A count can be simple

1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63, 64, 65, 66, 67, 68, 69, 70, 71, 72, 73, 74, 75, 76, 77, 78, 79, 80, 81, 82, 83, 84, 85, 86, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 100, 101, 102, 103, 104, 105, 106, 107, 108, 109, 110, 111, 112, 113, 114, 115, 116, 117, 118, 119, 120, 121, 122, 123, 124, 125, 126, 127, 128, 129, 130, 131, 132, 133, 134, 135, 136, 137, 138, 139, 140, 141, 142, 143, 144, 145, 146, 147, 148, 149, 150, 151, 152, 153, 154, 155, 156, 157, 158, 159, 160, 161, 162, 163, 164, 165, 166, 167, 168, 169, 170, 171, 172, 173, 174, 175, 176, 177, 178, 179, 180, 181, 182, 183, 184, 185, 186, 187, 188, 189, 190, 191, 192, 193, 194, 195, 196, 197, 198, 199, 200



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... or a count can be pretty darned complex

Variance on the estimator of the variance of a **Pacific cod** count based on ondeck observations of harvest in pots:

respectively. The expectation of Eq. 4.17 is
$$E\left({}_{1}\widehat{\boldsymbol{\Psi}}_{k}\right) = E_{s_{w}}\left\{\frac{M_{k}}{m_{2k}}\sum_{i=1}^{m_{2k}}E_{s_{L_{i}}}\left[\frac{N_{ki}}{n_{ki}}E_{s_{O}}\left(\frac{n_{k}}{n_{Ok}}E_{s_{A}}\left(\frac{N_{Om}}{n_{Am}}\psi_{Aki}\right)\right)\right]\right\}$$

$$= E_{s_{w}}\left\{\frac{M_{k}}{m_{2k}}\sum_{i=1}^{m_{2k}}E_{s_{L_{i}}}\left[\frac{N_{ki}}{n_{ki}}E_{s_{O}}\left(\frac{n_{k}}{n_{Ok}}\psi_{Oki}\right)\right]\right\}$$

$$= E_{s_{w}}\left\{\frac{M_{k}}{m_{2k}}\sum_{i=1}^{m_{2k}}E_{s_{L_{i}}}\left(\frac{N_{ki}}{n_{ki}}\psi_{Lki}\right)\right] = E_{s_{w}}\left(\frac{M_{k}}{m_{2k}}\sum_{i=1}^{m_{2k}}\Psi_{ki}\right) = \Psi_{k}.$$
The variance of the estimator can be written as
$$V\left({}_{1}\widehat{\Psi}_{k}\right) = \underbrace{V_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[V_{s_{O}}\left(E_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{L_{i}}}\left[E_{s_{O}}\left(V_{s_{A}}\left({}_{1}\widehat{\Psi}_{k}\right)\right]\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\left(V_{s_{A}}\right)\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\left(V_{s_{A}}\right)\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\left(V_{s_{A}}\right)\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\right)\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\right)\right\}}_{V_{i}} + \underbrace{E_{s_{w}}\left\{E_{s_{A}}\left(V_{s_{A}}\right)\right$$

$$\begin{split} V_{3} &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} E_{s_{L_{i}}} \left[V_{s_{O}} \left(\frac{\sum_{i=1}^{n_{2k}} N_{ki} \ n_{k}}{n_{ki} \ n_{Ok}} \psi_{Oki} \right) \right] \right\} \\ &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i=1}^{m_{2k}} E_{s_{L_{i}}} \left[\left(\frac{N_{ki}}{n_{ki}} \right)^{2} V_{s_{O}} \left(\frac{n_{k}}{n_{Ok}} \psi_{Oki} \right) \right] \right\} \\ &+ E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i=1}^{m_{2k}} E_{s_{L_{i}}} \left[\frac{N_{ki} \ N_{kj}}{n_{ki} \ n_{kj}} Cov_{s_{O}} \left(\frac{n_{k}}{n_{Ok}} \psi_{Oki}, \frac{n_{k}}{n_{Ok}} \psi_{Okj} \right) \right] \right\} \\ &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i\neq j}^{m_{2k}} E_{s_{L_{i}}} \left[\frac{N_{ki} \ N_{kj}}{n_{ki}} \frac{n_{k}}{n_{kj}} n_{k} \left(\frac{n_{k}}{n_{Ok}} - 1 \right) \frac{n_{k} \left[\operatorname{diag} \left(\mathbf{p}_{Lki} \right) - \mathbf{p}_{Lki} \mathbf{p}_{Lki}^{T}}{n_{k} - 1} \right] \right\} \\ &- E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i\neq j}^{m_{2k}} E_{s_{L_{i}}} \left[\frac{N_{ki} \ N_{kj}}{n_{ki}} \frac{n_{k}}{n_{kj}} \frac{n_{k}}{n_{k}} \frac{n_{k}}{n_{Ok}} - 1 \right) \frac{n_{k} \mathbf{p}_{Lki} \mathbf{p}_{Lkj}^{T}}{n_{k} - 1} \right] \right\} \\ &\text{and} \\ V_{4} &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} E_{s_{L_{i}}} \left[E_{s_{O}} \left(V_{s_{A}} \left(\frac{m_{2k}}{n_{ki}} \frac{N_{ki}}{n_{kj}} \frac{n_{k}}{n_{Ok}} \frac{N_{Om}}{n_{Am}} \psi_{Aki} \right) \right) \right] \right\} \\ &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} E_{s_{L_{i}}} \left[E_{s_{O}} \left(V_{s_{A}} \left(\frac{m_{2k}}{n_{ki}} \frac{N_{Om}}{n_{Ok}} \psi_{Aki} \right) \right) \right] \right\} \\ &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} E_{s_{L_{i}}} \left[\frac{m_{2k}}{n_{ki}} \frac{n_{k}}{n_{Ok}} \right)^{2} Cov_{s_{A}} \left(\frac{N_{Om}}{n_{Am}} \psi_{Aki} \right) \right\} \\ &+ \sum_{i \neq j}^{m_{2k}} E_{s_{O}} \left[\frac{N_{ki} \ N_{kj}}{n_{ki}} \left(\frac{n_{k}}{n_{Ok}} \right)^{2} Cov_{s_{A}} \left(\frac{N_{Om}}{n_{Am}} \psi_{Aki} \right) \right] \right\} \\ &= E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i=1}^{m_{2k}} E_{s_{L_{i}}} \left[\frac{N_{ki} \ n_{Ok}}{n_{ki}} \frac{n_{Ok}}{n_{Ok}} \left(\frac{N_{Om}}{n_{Am}} \right) \frac{N_{Om} - 1}{N_{Om} - 1} \right) \right] \right\} \\ \\ &- E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i=1}^{m_{2k}} E_{s_{L_{i}}} \left[\frac{N_{ki} \ n_{Ok}}{n_{ki} \ n_{Ok}} \left(\frac{N_{Om}}{n_{Am}} \right) \frac{N_{Om}}{n_{Am}} - 1 \right) \frac{Po_{ki} P_{Oki}^{T}}{N_{Om} - 1} \right] \right\} \\ \\ &- E_{s_{w}} \left\{ \left(\frac{M_{k}}{m_{2k}} \right)^{2} \sum_{i=1}^{m_{2k}} E_{s_{L_{i$$

An observation



Counting **fish** is just like counting **trees**, except they're *invisible* and they *move* ...

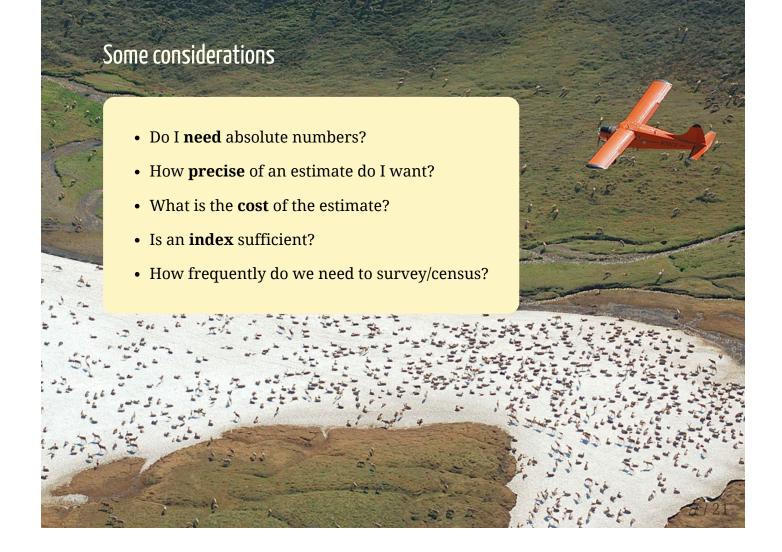


Three broad approaches (with sub-categories)

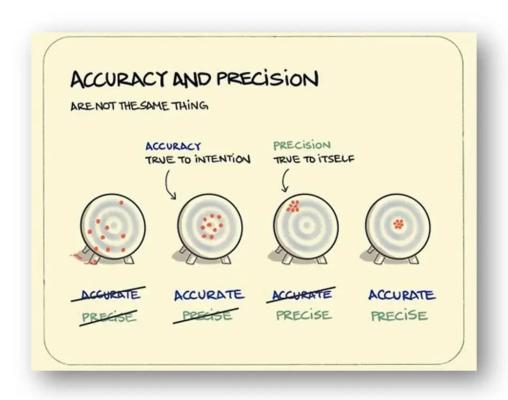
1. Compete census (total count)
2. Sample count

• Counts along transects or variable plots
3. Mark-racapture/resight
4. Population index

• count signs or correlates, not animals



Two important considerations:



Accuracy

Is the estimate **biased**?

• On average, on target

Determined by **design**.

how well you throw the dart?

Can be difficult to assess.

beacuse there isn't usually a dartboard!

Precision

What is the *error* or *variance* or *spread* on the resulting estimate?

Quantified with Confidence Intervals (C.I.) (or coefficients of variation (C.V.))

ACCURACY AND PRECISION

ACCURACY TRUE TO INTENTION

ACCURATE

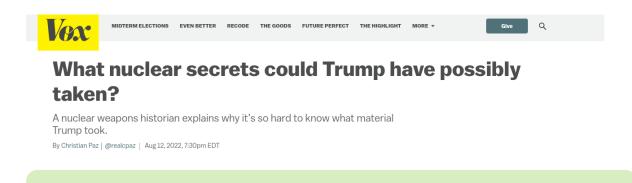
ARE NOT THE SAME THING

Determined by **effort** and computed with (sometimes very fancy) **statistics**.

Generally: bigger the **sample** = smaller **error** = higher **precision**.

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Very accurate, but very imprecise:



"It could be anything ranging from something that would endanger the lives of hundreds of millions of people to something that has no impact on anything whatsoever. That's how vague the classified categorization is," Alex Wellerstein, a historian of science and nuclear weapons, told me.

General Goal of Abudance Estimation

Increase both **accuracy** and **precision** as contraints on **effort** (& costs).

Generally, there a higher premium on accuracy

• (i.e. better an unbiased but imprecise estimate, than a highly precise but biased estimate).

When might BIAS not be so important?

If the bias is consistent, repeated measures can tell you how things are *changing*

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Method I: Total Count, aka. Census

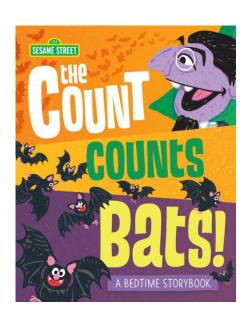
Pros

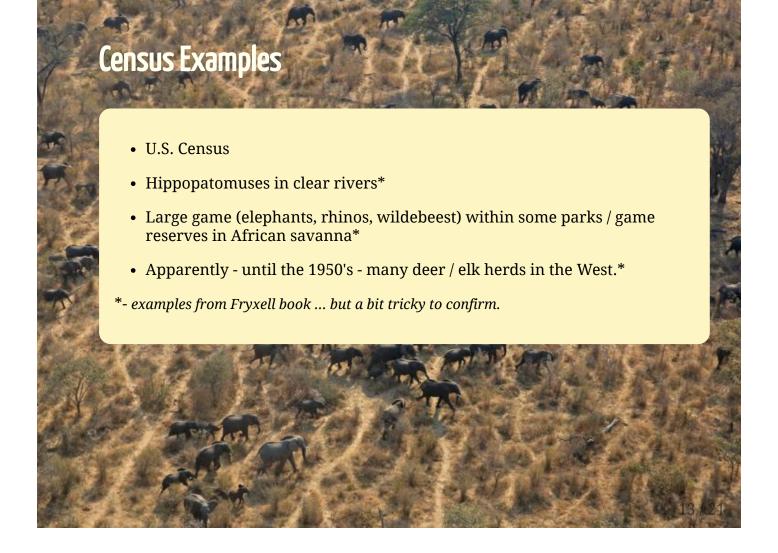
- Simple to explain!
- Simple math (arithmetic)!
- Very precise

Cons

- Usually VERY difficult / expensive to perform
- Only possible for certain kinds of animals
- · Almost always biased!

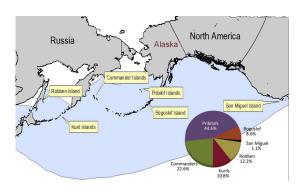
What kinds of animals can we census?





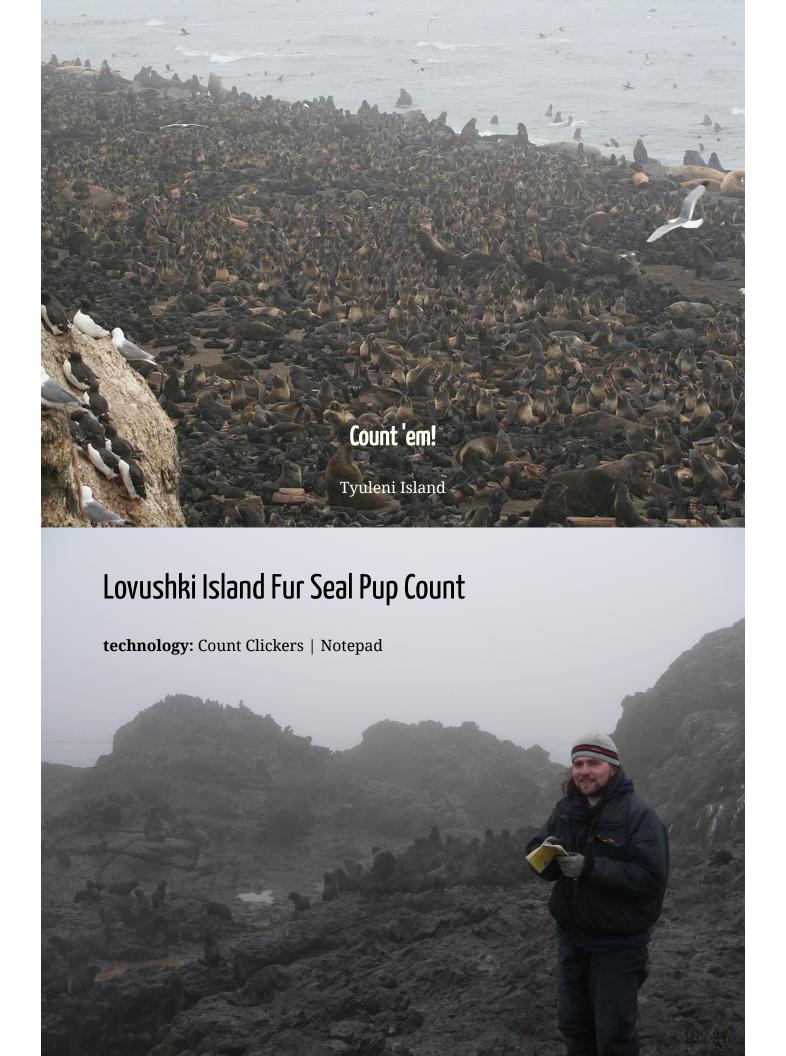
Northern Fur Seals Callorhinus ursinus





- Once externely abundant
- VERY heavily harvested
- Paid off 1867 purchase Alaska in 30 years
- Reproduce (essentially) in only 6 rookeries worldwide
- At heart of the first international wildlife management treaty.

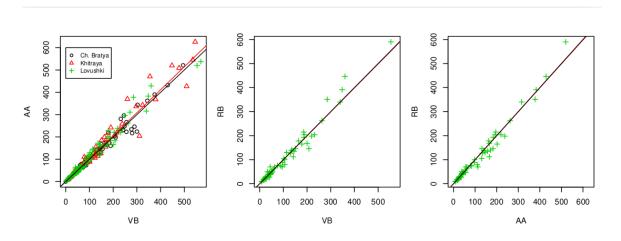






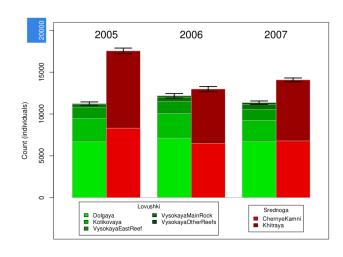
Fur seal count: Source of variation?

Individual counters



Pretty good agreement.

Fur seal count: High Precision!



Point Estimate:

$$\widehat{N}=28,792$$

Standard Error (s.e.):

$$\sigma_e=216$$

95% Confidence Interval

$$\widehat{N}\,\pm 2\sigma_e = (28,630-29,220)$$

Coefficient of variation

$$rac{\sigma_e}{\widehat{N}}=0.75\%$$

